

Schumpeterian disaggregation and integrated assessment: An endogenous, stock-flow consistent economy in disequilibrium for FRIDA v2.1

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Abstract. Integrated assessments of climate change require models capable of capturing the coupled dynamics of natural and socioeconomic systems. This paper presents the economy module of FRIDA v2.1, a Schumpeterian, disequilibrium framework of endogenous growth designed to address several limitations of contemporary integrated assessment models (IAMs). The module incorporates monetary, financial and innovation dynamics driving productivity and endogenous business cycles, while allowing explicit representation of how climate impacts propagate through various institutional sectors and economic processes. Its process-based structure replaces aggregated damage functions with disaggregated, empirically grounded mechanisms, improving the traceability of assumptions and enabling the study of climate-finance interactions—including risks of disorderly transitions—absent from mainstream IAMs. Calibration against historical data demonstrates that the model produces dynamics broadly consistent with observed macroeconomic developments over the period 1980–2023. A 100,000-member ensemble simulation communicates the uncertainty in projections through 2150 while revealing endogenous constraints on economic activity. We show that without further action to combat climate change, expected climate impacts not only affect economic production, primarily through reduced investment growth and financial fragility, but also government budgets which come under stress owing to the increasing burdens of unemployment and demographic change. By providing a transparent, modifiable platform for simulating monetary, financial, and innovation dynamics under climate constraints, FRIDA v2.1 expands the analytical scope of IAMs and supports richer exploration of transition pathways.

1 Introduction and state of the art

Integrated assessment models (IAMs) are important tools for shaping global climate policy. At the international level, they feature heavily in the IPCC reports (IPCC, 2023) and inform climate negotiations under the UNFCCC (Science in the UNFCCC negotiations, 2025). In the United States, they are used to calculate carbon prices for policy appraisal (Rennert et al., 2022; Stern et al., 2022). In the European Union, model-based evidence of macroeconomic impacts has become a near prerequisite for accepting new climate or energy policy proposals (Pollitt and Mercure, 2018). IAMs' widespread use and growing influence have naturally invited scrutiny from the research community, leading to the identification of several limitations. It has been suggested that widening the underlying methodological scope is needed to overcome these limitations (Donges et al., 2021; Keppo et al., 2021; Stern et al., 2022). The FRIDA v2.1 model represents one such effort to broaden the methodological foundations of integrated assessment. The purpose of this paper is to document the economy module built for FRIDA v2.1 and its contributions to addressing some of these limitations, which are described below.

Most IAMs rely on highly aggregated damage functions, translating the complex relationships between temperature increases and economic consequences into a direct reduction in GDP. These functions' empirical foundations are limited because of data constraints, making their parametrisation highly uncertain (Farmer et al., 2015; Pindyck, 2013, 2017). Model outcomes, including policy-relevant optimal emission trajectories and the social cost of carbon estimates, demonstrate

extraordinary sensitivity to damage function specifications (Gillingham et al., 2018; Hänsel et al., 2020; Nordhaus, 2019). Furthermore, owing to their aggregate nature, these damage functions operate as statistical black boxes (Pindyck, 2017). This aggregation necessarily compromises the traceability of model assumptions, making it impossible to identify how climate damage disrupts specific macroeconomic systems and sectors. The resulting opacity in model structure constrains researchers' ability to test individual assumptions and validate specific climate damage mechanisms (Elster, 2015; Meadows and Wright, 2008).

Similarly, most IAMs feature heavily aggregated economy modules. Representations of the financial system, notably, are either absent or omit the complex dynamics of credit creation, financial intermediation, and monetary policy, despite substantial literature documenting the financial risks posed by climate change (Battiston et al., 2021a; Pollitt and Mercure, 2018). Researchers have described how climate change and resulting extreme weather events pose risks to the global financial system, ranging from physical damage to financial assets to transition risks from stranded assets (Battiston et al., 2017, 2021b; Lamperti et al., 2019; Mandel et al., 2025). The mounting evidence has made financial and monetary authorities increasingly attentive to climate-related risks (Carney, 2015; Kiley, 2021; NGFS, 2019) and even prompted amendments to financial stability policies (Brunetti et al., 2021; Giuzio et al., 2019), making the inclusion of finance in IAMs timely.

In addition to understanding the vulnerability of financial systems to climate impacts, researchers call for models that are able to explore how global finance is poised to mitigate climate change. Sanders et al. (2022) noted that IAMs without finance leave “a crucial gap” in our ability to conduct macro model-based analyses of climate policy. A disorderly transition—where delayed recognition of climate risk leads to sudden, tardy action—is probable and poses systemic risks. Such a green transition could beget financial instability (Battiston et al., 2017, 2021a; Carattini et al., 2023; Garcia-Jorcano and Sanchis-Marco, 2025; Ojea-Ferreiro et al., 2024). Both Sanders et al. (2022) and Stern et al. (2022) warned that a disorderly transition could involve large and rapid changes in the price of carbon, resulting in sudden changes in asset values, triggering system-wide financial distress. These volatile dynamics are outside the scope of most contemporary IAMs because of their equilibrium assumptions and omission of finance, which leads us to the next IAM limitation: their lack of business cycle dynamics and short-term phenomena with long-term consequences.

In the most comprehensive treatment of the topic to date, Annicchiarico et al. (2022) noted that integrating business cycles in climate policy analyses represents a frontier area where substantial research gaps remain. Empirical evidence supports the hypothesis that emissions are procyclical (Doda, 2014). This finding indicates that instruments such as a carbon tax may be more effective if adjusted to the business cycle (Annicchiarico et al., 2022). Most contemporary IAMs are ill suited for evaluating the dynamic relationship between business cycles and climate policy (Stern et al., 2022). Adequately modelling these dynamics requires the integration of short-term mechanisms within a modelling framework that has traditionally favoured a long-term perspective (Pollitt and Mercure, 2018).

Neglecting short-term volatile mechanisms has implications beyond policy assessment. For example, extreme climate events can affect both existing economic activities and planned investments (Griffin et al., 2019). This is borne out in the growing body of research documenting the financial impact of climate change, especially in regard to failure rates of investments, leading to increased bankruptcies (Bartsch et al., 2024; Carattini et al., 2023; Feng et al., 2024). These unexpected defaults produce short-term negative employment shocks and drive long-term changes in lending standards, as the financial sector adapts its future growth expectations accordingly, negatively affecting investment and therefore both quantitative growth and qualitative economic development (Dell’Ariccia and Marquez, 2006; Fishman et al., 2024; Lown and Morgan, 2006; Rodano et al., 2018).

A more realistic representation of financial and monetary mechanisms and their dynamics further supports the endogenous modelling of innovation. Stern et al. (2016) emphasised that most IAMs fail to capture the feedback loops in innovation processes, particularly the interactions across the economy that drive institutional and behavioural change. Mercure et al. (2019) extended this critique, showing that outcomes in climate–economy models are vastly different depending on how innovation is represented. They stress that innovation cannot be treated as an exogenous cost-reducing trend, as this approach fails to take into account the disruptive nature of the innovation process and its multiple feedbacks (Antonelli, 2017; Schumpeter, 1934) but should instead be endogenised within a framework that considers financial conditions, policy interventions, and other institutional dynamics—another direct call for economic disaggregation.

To address the highlighted gaps, the economy module built for FRIDA v2.1 rests on the following conceptual foundations: it is built on Schumpeterian theory, it operates in disequilibrium, all growth components are endogenous, and it

features a stock-flow consistent financial architecture. These characteristics, elaborated on in Section 2, allow FRIDA v2.1 to analyse the two-way interactions of climate change through a multitude of direct and explicit impacts documented in Section 3. This opens new possibilities for understanding climate–economy interactions and evaluating policy interventions.

Table C1 in Appendix C provides an overview of IAMs with similar aims, alongside the classical DICE model for reference. While other models address some of the research gaps described above, they typically concentrate on the economy in isolation. FRIDA v2.1 stands out by including a similar level of detail in the other domains of the human–earth system (Schoenberg et al., 2025b), and is one of only two models with fully endogenous innovation.

FRIDA v2.1, and its novel economic module documented in this paper, aims to contribute to the recent research stream addressing these limitations. One body of work particularly influential for the FRIDA project is the “coupled human and natural systems” (CHANS) modelling approach, which emphasises that human and natural systems are defined by feedbacks, thresholds, nonlinearities, time lags, and emergence and link flows of matter, energy, and information (Alberti et al., 2011; Kramer et al., 2017; Liu et al., 2007). FRIDA v2.1 takes on this approach by disaggregating climate damage functions into multiple categories (see Wells et al. (2025) for a detailed treatment of the biophysical processes behind climate losses and damage, as represented in FRIDA v2.1), breaking down aggregate climate impacts into tangible, observable processes that can be empirically parameterised—processes that an aggregate damage function necessarily obscures. FRIDA v2.1’s economy module adopts a disaggregated process-based approach to socioeconomic modelling that integrates short-term monetary, financial and innovation dynamics in the analysis.

This documentation paper is structured as follows. Section 2 expands on the module’s key conceptual foundations by establishing the theoretical rationale underlying model design choices and its attending limitations. Section 3 provides a detailed description of the module’s structure across seven interconnected submodules, complete with stock and flow diagrams. Section 4 describes the calibration methodology and our protocol for the treatment of uncertainty. Section 5 presents simulation results demonstrating the model’s performance against historical data and projection confidence intervals that explicitly communicate the project’s inherent uncertainty. Section 6 concludes the paper with a run-through of contributions and a discussion of future applications.

2 Analytical aims, conceptual foundations, and limitations

2.2 Analytical aims

FRIDA v2.1 is a multipurpose model for climate–economy analyses and policy experimentation. Its analytical aims are to close feedback loops across human, economic, and climate systems in a process-based manner and to identify the most essential feedback processes in a climate context (Schoenberg et al., 2025b). These aims require certain considerations in the development of the model. The first consideration is the representation of multiple climate impact pathways affecting specific economic processes and actors, rather than a single aggregate damage function. The second consideration is the need to include finance and monetary mechanisms of climate damage and adaptation. The third consideration is the need to capture emergent dynamics from interactions between short-term dynamics with long-term consequences, such as financial fragility, and long-term dynamics susceptible to short-term disruption, such as innovation. The fourth consideration is to endogenously model aspects of human behavioural change relevant to the climate context. The final and fifth consideration is to enable fully endogenous simulation of all of these components without the assumption of an exogenous “social decision maker” driving investment and mitigation choices.

2.3 Conceptual foundations

To achieve these analytical aims and satisfy the related requirements, the economy module of FRIDA v2.1 has been developed as a Schumpeterian disequilibrium model of endogenous growth (Antonelli, 2017; Dosi et al., 2010; Schumpeter, 1939). The Schumpeterian framework provides a consistent understanding of financial and innovation dynamics. It operates in disequilibrium to enable the analytical integration of short-term and long-term dynamics and their interactions. Growth and innovation are treated as endogenous to evaluate the consequences of climate damage and climate policy for

135 productivity growth and the opportunities offered by productivity growth for mitigation and adaptation purposes. The following paragraphs elaborate on these characteristics in turn.

While it is possible to include financial mechanisms in IAMs on the basis of a real analysis approach (see Adelman & Yeldan (2000) for an exogenous solution), it would have excluded the modelling of endogenous short-term systemic crises with enduring consequences. Owing to the potential relevance of these mechanisms for the analysis of the economic consequences of climate change, we opted to integrate them within the context of a monetary, stock-flow-consistent (SFC) approach, as suggested by Pollitt and Mercure (2018). We further adopted Keppo et al.'s (2021) recommendations to track debt accumulation, borrowing capacity, and interest rates to assess the potential disruptive effects of climate risks and transitions. This has the further benefit of enabling FRIDA to evaluate the inflationary consequences of both climate damage and climate policy.

145 To represent disequilibrium dynamics, the classical aggregate production function is replaced by a dynamic circular income flow framework, depicting a disaggregated ensemble of firms, owners and workers (Schumpeter, 1934). Eschewing the equilibrium approach and its inbuilt assumptions is required in light of the idiosyncrasies affecting the economic analysis of climate change. The significant uncertainty affecting climate change dynamics, the global nature of the problem, and its complex social and distributional aspects imply significant associated market failures (Stern et al., 2022). Even if market instruments could work perfectly, these instruments are either in their infancy or simply missing (Stern, 2022). While methodologically convenient, equilibrium assumptions are inconsistent with the defining characteristics of the analytical issue at stake. Since the process-based methodology adopted for FRIDA 2.1 allows for the construction of disequilibrium models (Cavana, 2021), we developed FRIDA 2.1 accordingly.

Endogenous growth is a requirement for any model aiming to provide a viable environment for policy analysis. The Schumpeterian approach to the analysis of growth focuses on the role played by innovative investments in expanding the potential production frontier. This enables us to endogenise both the processes of growth through accumulation and the processes of growth achieved through qualitative changes in the use of economic resources. Furthermore, Schumpeterian analysis stresses the potential short-term negative effects induced by innovative efforts (Aghion et al., 2012, 2014; Quatraro, 2016; Schumpeter, 1939), thus introducing an additional analytical dimension expanding policy analysis beyond the traditional contraposition of growth and environmental sustainability. This is in line with FRIDA's primary object of analysis: A system-wide process of transition, successful or otherwise, towards more sustainable development pathways brought about by increasing climate-related constraints. This is an inherently innovative process, implying qualitative, potentially disruptive change on a global scale involving both private and public agents. Consequently, the Schumpeterian framework, focused on innovation and qualitative development, private and public, including both positive and negative consequences, is an appropriate theoretical instrument for the work at hand (Antonelli, 2017; Schumpeter, 1942).

165 2.3 Limitations

While the economy module of FRIDA v2.1 addresses some limitations of current IAMs, it also shares others and introduces new ones inherent to its approach. The main limitations concern global aggregation, optimal-policy identification under disequilibrium, and the consequences of the business cycles dynamics emerging from the Schumpeterian framework.

170 The first limitation concerns geographic aggregation. FRIDA v2.1's global aggregation masks local climate impacts and precludes differentiated, locally targeted policy experimentation. This is a familiar limitation of globally aggregated IAMs and one we accept as a necessary consequence of the model's scope.

The second limitation concerns optimal policy identification. In the published configuration the model does not calculate a utility or welfare measure. While this could be added (e.g., based on consumption measures), the optimisation of policy measures is not implemented in the model setup and is not required for running the model. Policy choices are an expected user input rather than optimised based on an assumed welfare function. While many IAMs are designed for policy optimisation, Ackerman et al. (2009) note that the resulting optima are highly sensitive to assumptions regarding discounting, technological change, and uncertainty. We therefore eschew formally optimal policies for an approach that avoids conditioning policy analysis on these contested assumptions. Considering our aims, and the scope afforded by our approach, we consider this trade-off acceptable. Nonetheless, it would be feasible to embed the model in a policy optimisation harness where an external optimiser can vary the policy levers intended for user input automatically to find a combination maximizing a payoff calculated based on model outputs.

185 The third limitation arises from the model's endogenous business cycles. As a consequence of its Schumpeterian foundations (Schumpeter, 1939), FRIDA v2.1 features emergent business cycles from the interaction between innovation and finance, a relatively unique feature among current IAMs of comparable scope. While the dynamic is ultimately nested in production dynamics, their magnitude and timing are affected by financial dynamics and aggregate financial actors. (Aghion et al., 2014; Schumpeter, 1939) The cycles featured in the model, however, are significantly different from actual business cycles. The former are produced by a single, endogenous driver, namely the performance of innovative activities in a market economy, while the latter are complex phenomena driven by multiple endogenous and exogenous drivers, and their interactions (Ramey, 2016). The representation of realistic business cycles lie outside the model's scope. The model is
190 therefore unable to either accurately reproduce specific historical cycles, or forecast the timing and amplitude of future ones. The model's cycles should be read as illustrating the qualitative behaviour of one structural mechanism, rather than as point predictions of macroeconomic activity. Their purpose, instead, is to enable the analysis of climate change, and climate-related policies, with innovation-related endogenous business cycle dynamics.

195 The inclusion of cycles within FRIDA also produces a calibration trade-off. The historical calibration data reflects the full gamut of both endogenous drivers and exogenous shocks (such as pandemics and wars) that lie outside FRIDA v2.1's scope. Consequently, calibration attributes the full variance generated by these complex processes and events to the single Schumpeterian mechanism present in the model, inevitably exaggerating its relative strength. To mitigate this issue, FRIDA v2.1 places uncertainty quantification at the forefront of its methodology rather than obscuring it. The precise steps in this uncertainty protocol are detailed in Section 4.

200 3 Model structure and submodules

The Economy module is one of seven modules in FRIDA and represents the global economy. All internal processes are simulated in nominal terms, which is consistent with the module's monetary framework (real GDP is calculated by adjusting for inflation in the GDP submodule, as described in Section 3.6). It receives inputs from the Climate, Land Use and Agriculture, Demographics, and Energy modules and provides outputs to the Demographics, Resources, Energy and
205 Behavioural Change modules. A schematic representation of these interactions is provided in Fig. 1. While embedded in FRIDA, the Economy module can also function independently as a stand-alone macroeconomic model of growth under exogenous environmental constraints. To explain the high-level interactions between all the modules and how the Economy module influences the rest of the model, see Schoenberg et al. (2025b).

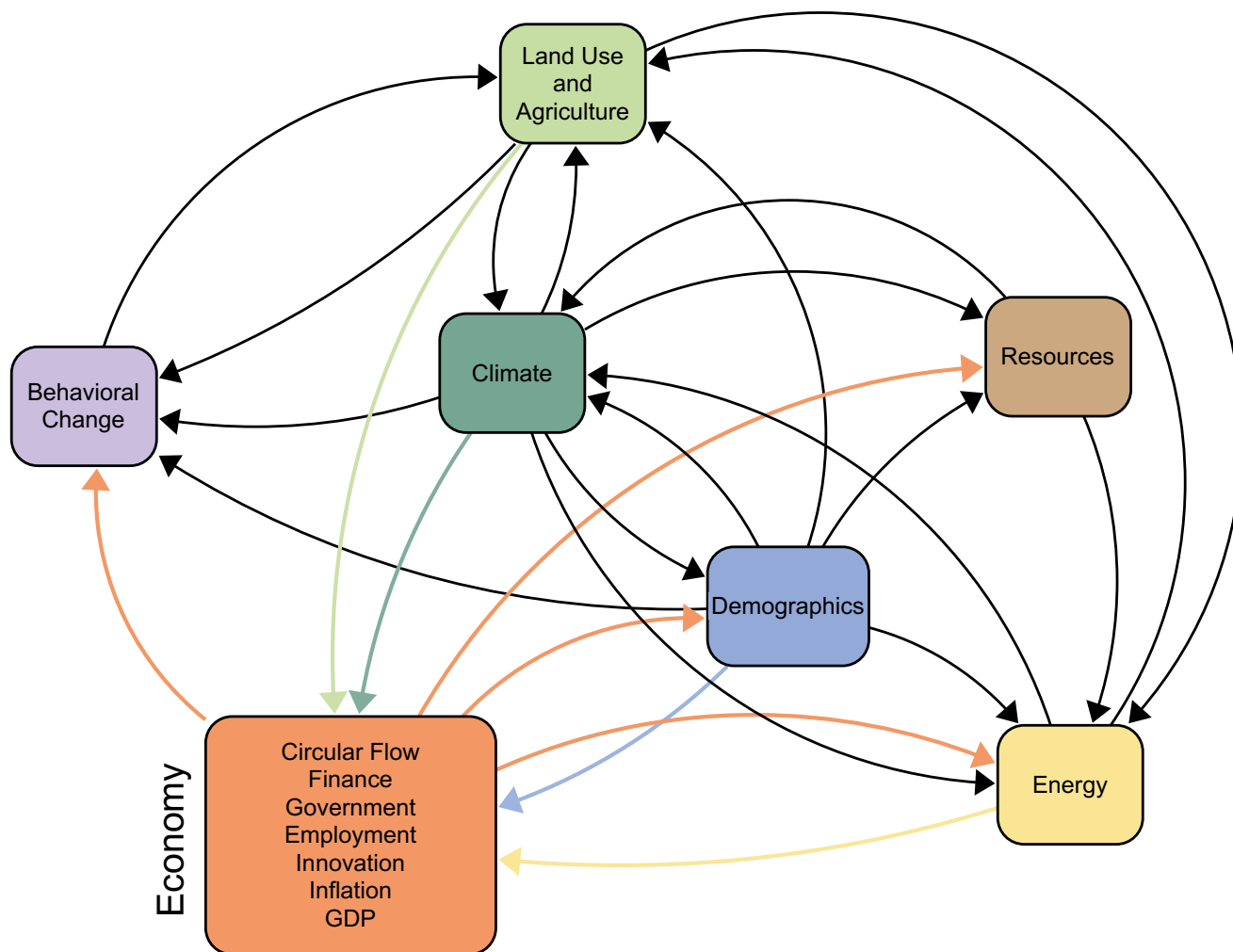


Figure 1: The seven high-level modules comprising the FRIDA model, the connections between them, and the Economy module's component submodules.

The Economy module includes 761 equations, of which 33 are state variables. It is calibrated against data sourced for the period 1980–2023, covering the key macroeconomic indicators GDP, investment, consumption, government expenditure, the government debt-to-GDP ratio, inflation, unemployment, and wages. These time series were sourced from the World Bank, IMF, OECD, WID, and ILO, ensuring consistency and reliability (see Appendix A for the specific payoff elements and dataset references).

The internal structure of the Economy module consists of the seven submodules listed in the Economy module box in Fig. 1, each representing distinct but highly interconnected processes. Before turning to the detailed documentation, we provide a brief description of the seven submodules and their roles in the module's feedback structure. The Circular Flow submodule simulates firms' production and households' consumption. The Finance submodule mediates private investment affecting the flow of funds in and out of the circular flow, employment levels and interest rates. The Innovation submodule, together with the Finance submodule, governs R&D orientation and exploratory lending, determining the productivity growth that drives long-term economic development while displacing existing investments and workers in the short term. The Government submodule sets fiscal and monetary policy, with tax revenue, transfers, public debt, and the policy rate responding to conditions elsewhere in the economy. The Employment submodule translates investment flows into labour

demand, wage formation, and unemployment, feeding back into consumption and government transfers. The GDP and Inflation submodules aggregate these flows into nominal and real output and determine the price level, which in turn affects consumption, wages, and the policy rate. This section documents the processes represented within each of these submodules and how they interact in more detail. First, the Circular Flow submodule is described, followed by the Finance, Innovation, Government, and Employment submodules, and finally the GDP and Inflation submodules.

3.1 Circular Flow

The Circular Flow submodule is used to simulate the processes of production and consumption and represents the behaviour of two aggregate agents: households and firms. Fig. 2 showcases the submodule’s main stocks and the flows between them. The processes represented in the government, employment, and finance modules greatly affect the circular flow but are modelled separately for conceptual clarity; the details of these mechanisms are covered in the sections corresponding to their respective submodules.

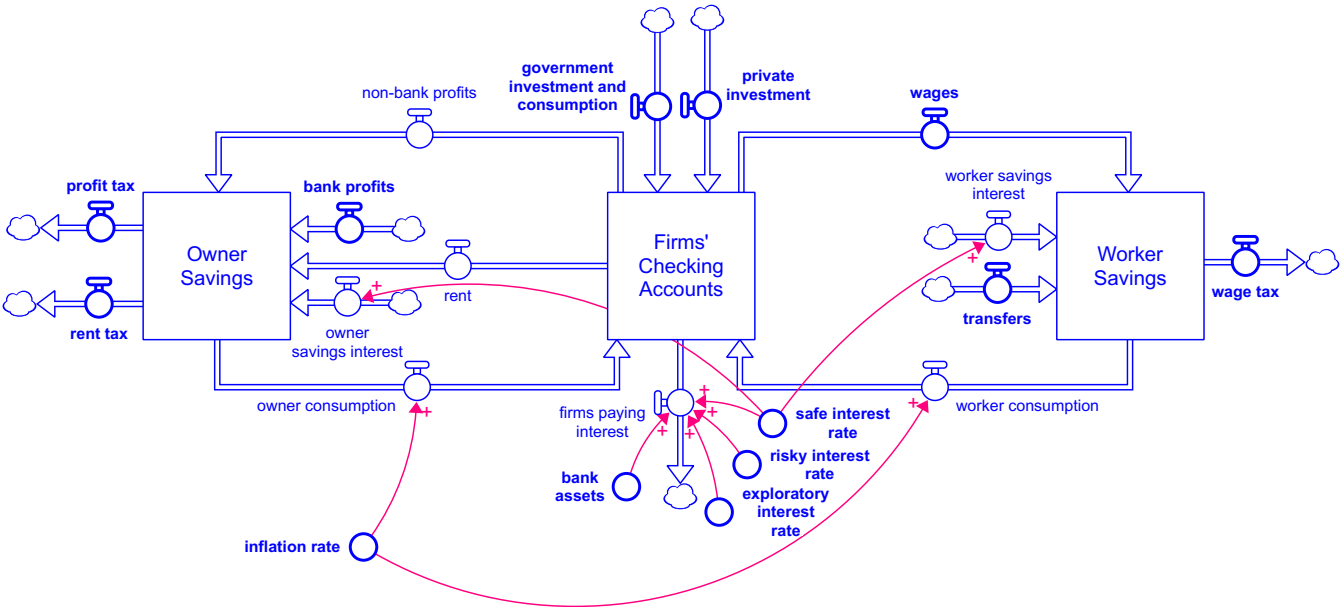


Figure 2: A simplified stock and flow diagram of the Circular Flow submodule. Bold elements receive input from elsewhere in the model.

Household money flows are disaggregated into those originating from the labour supply aspect and those related to the ownership aspect of households, following established practices in stock-flow consistent modelling of a monetary economy (Godley and Lavoie, 2007). Fig. 2 shows how all income related to ownership flows into owner savings, while labour income flows into worker savings, thereby tracking the mechanisms affecting the flow and accumulation of different income streams separately. A percentage of both the owners’ and workers’ net income is spent on consumption. Additional costs of ownership stemming from climate change in the form of costs related to retreating from sea level rise (SLR) are included in owner consumption (see Ramme et al. (2025) for documentation of FRIDA’s Sea Level Rise Impacts and Adaptation module). The remaining income after consumption accumulates as savings, represented by two stocks, owner savings and worker savings, and conceptualised as checking accounts in the aggregate banking system, representing aggregate deposits from owner and worker incomes, respectively.

The percentage of income spent on consumption is calculated using the same method for workers and owners but parameterised differently under the assumption that a larger share of worker income is spent immediately, while a larger share of owner income is saved. This is because asset-based income by and large tends to be allocated to long-term savings goals, including retirement, while a higher portion of income from labour tends to be spent on day-to-day expenses (Carroll

et al., 2017). The level of consumption of both groups is determined by the relationship between income and each group's dynamic savings goal. This savings goal, which represents a desired level of savings relative to income, adjusts gradually over time as each group's financial situation changes. When income increases, both groups revise their savings target upwards after a delay, reflecting a gradual reassessment of their wealth (Jappelli and Pistaferri, 2010). With a similar lag, the savings goal adjusts downwards with reductions in aggregate income. The consumption decision follows a rule where a fraction of each group's income, minus the shortfall between their current savings and their target savings level, is spent on consumption—with a hard floor to prevent households from having unrealistically low consumption levels under extreme conditions (Andrejeva et al., 2010). Inflation has an immediate effect on the consumption of both groups; as prices change, net of the negative impact of additional unemployment, consumption increases in an attempt to maintain the same standard of living (Burke and Ozdagli, 2023).

The consumption expenditure of each group flows into a third state variable, the firms' checking accounts. This variable, shown in figure 2, represents all the deposit accounts for all the world's firms. The income streams that flow from firms checking accounts to households consist of wages, rent, and profits. The processes that determine the cost of wages and rent are modelled in the Employment submodule and discussed in Section 3.5. Together, wages and rent represent the compensation received by households in exchange for their contribution of productive services to the production process. These payments are drawn from the firms' checking accounts. Profits consist of the residual firm income after all the required payments have been made by firms. They are distributed to owners gradually as firms manage outflows to maintain the required working capital (Larkin et al., 2017).

We represent government transfers (welfare payments) as contributing to the worker subset of household income. The process for determining government transfers is modelled in the government module and discussed in Section 3.4. The savings of both workers and owners receive interest from the banks through a process described in the Finance submodule (Section 3.2). Fig. 2 shows that bank profits are represented separately from the profits already discussed. Bank profits are not drawn from the firms' checking accounts because banks are modelled separately from general firms. This process is explained in Section 3.2. Moreover, wages, profits and rent are taxed, creating revenue for the public sector, which is used in the Government submodule (Section 3.4.).

The circular flow, as explained thus far, cannot grow on its own. Income would circulate between firms and households, leading to a stable equilibrium as savings goals are met (Schumpeter, 1934). Growth is achieved through private investments originating in the Finance submodule (Section 3.2.), which lends money to firms, enabling them to increase the scale of their productive activities. Firms pay interest on these loans depending on the risk level as determined by the Finance submodule (Section 3.2). A second potential source of growth of the circular flow is government expenditures; when government expenditures exceed tax revenues, they increase the general income level, although the increase may be purely nominal depending on productivity growth and the amount of slack present in the economy. These government expenditures are determined in the Government submodule (Section 3.4). The next section describes the Finance module—how it mediates private investment and enables growth.

3.2 Finance

The Finance submodule is modelled as a single aggregate financial intermediary. It simulates mechanisms governing private investment, credit creation, allocation of resources to exploratory lending, and risk management. Key concepts include loans, which are classified as performing¹, nonperforming, safe or exploratory, lending standards, which represent banks' attitudes towards lending risk, and defaults, which arise from the eventual failure to repay nonperforming loans.

¹ As shorthand, performing and nonperforming loans are represented in the model and Fig. 3 as “Good Loans” and “Bad Loans,” respectively.

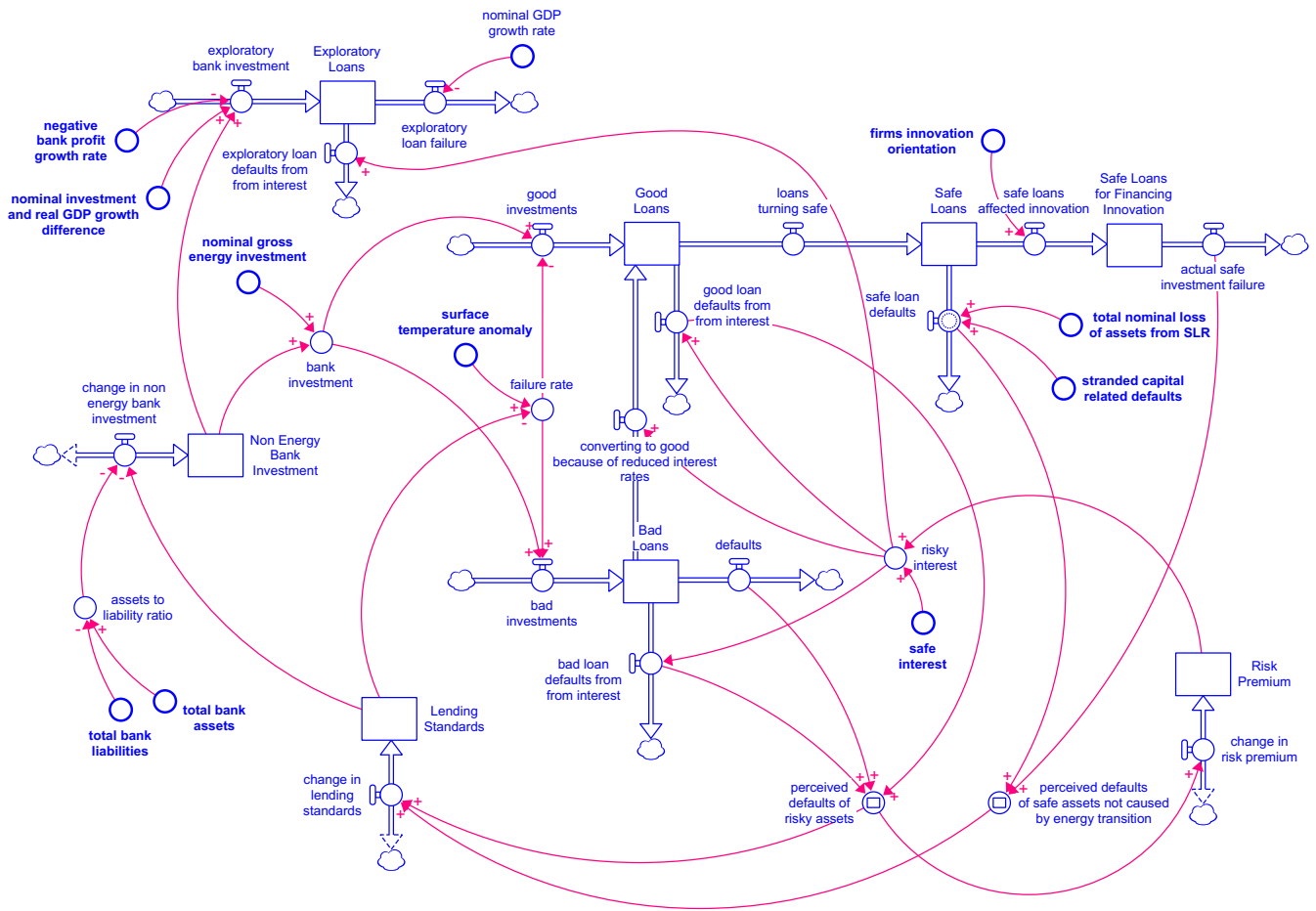


Figure 3: A simplified stock and flow diagram of the finance sector. Bold elements receive input from elsewhere in the model.

3.2.1 Bank profits, solvency, and government intervention

Households' accumulation of savings and the banking sector's profit incentives drive the issuance of loans, as banks attempt to generate a sufficient amount of interest-paying assets to remunerate deposits and distribute profits, a monetary rendition of the savings/investments nexus (Bernanke and Blinder, 1988). The issuance of loans expands both sides of the banking sector's balance sheet: when banks extend credit, it increases their assets in the form of loans (represented as exploratory, good, bad and safe loans in Fig. 3) and creates new liabilities in the form of additional deposits held by firms ("Firms' Checking Accounts" in Fig. 2) (McLeay et al., 2014)². This mechanism injects new money into firms, from where it circulates between households' and firms' deposit accounts through the remuneration of productive services and consumption as described in Section 3.1. The difference between accumulated loans in the Finance module and the accumulated deposits in the Circular Flow constitutes the banks' equity.

The amount of profits distributed by banks to owners in the Circular Flow submodule (Section 3.1) is a fraction of the total equity and is dynamically adjusted on the basis of the annual growth rate of equity. Declining (rising) equity puts downwards (upwards) pressure on bank profit distribution (Menicucci and Paolucci, 2016). Bank profits flow into the circular flow as an income stream. If the bank asset-to-liability ratio falls below an exogenous threshold, we simulate bailout

² FRIDA v2.1 does not model consumption credit.

policies through which the government intervenes and provides solvency support, subsidising the banking sector to ensure financial stability (Gup, 2003).

3.2.2 Investment, lending and loan classification

315 The banking sector aims to generate profits. To do so, they need to generate performing loans, expanding their balance
sheets in the process. The banks' desire to expand their asset base is modelled as a long-term historical reference of the past
rate of growth in loan volume (Rodano et al., 2018), adjusted for their equity target. Fig. 3 shows how a ratio of assets to
liabilities is an input to bank investment: when the ratio is below a desired threshold that puts upwards pressure on the
banks' desired investment growth rate, and vice versa. The preponderance of bank lending is governed by mechanisms
320 interior to the Finance submodule, but those investments pertaining to energy are determined by FRIDA's energy module, as
documented in Schoenberg et al. (2025b).

This quantitative expansion drive is tempered by the desire to minimise the origination of nonperforming loans. The
primary instrument to achieve this aim is the screening of potential borrowers and investment opportunities, represented by
lending standards. Lending standards, a stock in Fig. 3, are represented as a relative index adjusted over time in response to
325 the default rate, which is the realised defaults relative to the loan portfolio of performing and nonperforming loans, and past
lending standards (Fishman et al., 2024; Lown and Morgan, 2006). If the rate of loan creation exceeds the economy's growth
potential, loan quality deteriorates, leading to increased defaults (Kraft and Jankov, 2005). As the default rate increases,
lending standards increase, which reduces lending growth, and vice versa (Gjeçi et al., 2023).

In the model, we classify new loans as "performing", "nonperforming", or "exploratory" at issuance and
330 accumulate them into distinct stocks as shown in fig. 3. "Exploratory" loans stand for funding of innovative projects with the
possibility of contributing to productivity growth. The origination of exploratory loans is modelled separately from that of
conventional performing and nonperforming loans and is described together with innovation and productivity in Section 3.3.

Owing to information asymmetry, banks do not know which new loans are going to be nonperforming, but they do
know that some of them will. Consequently, the banks consider all new lending as risky and apply risk premia to newly
335 originated loans (Liao et al., 2009) (see Section 3.4.2 for interest rate formation). As these risky loans mature, nonperforming
loans default and are written off as losses for the bank. On the other hand, performing loans are reclassified over time as safe
loans. Once a loan has been demonstrated to be safe, a risk premium is no longer applied. Exploratory loans, however,
maintain their risk premium.

The proportion of new conventional loans classified as nonperforming at origination is determined by the "failure
340 rate". As shown in Fig. 3, the failure rate influences whether new investments by the banks are "good" or "bad", where good
bank investments accumulate as performing loans, and bad investments accumulate as nonperforming loans. The failure rate
has a calibrated baseline, reflecting normal liquidity conditions with well-functioning financial markets, and is dynamically
adjusted on the basis of economic and climatic conditions (World Bank, 2023e). Higher lending standards lead to stricter
screening of investment opportunities, reducing the failure rate as banks allocate credit more selectively (Lown and Morgan,
345 2006; van der Veer and Hoeberichts, 2016). Conversely, looser lending standards increase the failure rate as riskier
investments are pursued (Dell'Ariccia and Marquez, 2006; Rodano et al., 2018). Additionally, when the growth rate of
annual investment volume exceeds GDP growth, the failure rate also increases. This occurs because investment expands
faster than the economy is able to generate profitable investment opportunities (Beck et al., 2015; Kraft and Jankov, 2005).
However, this mechanism is not visualised in Fig. 3 as a measure to simplify the diagram. Finally, global climate change,
350 represented by increases in surface temperature anomaly (STA), further increases risk and is shown in Fig. 3 as an input to
the failure rate. As global temperatures increase above preindustrial levels, extreme weather events increase the probability
of new investments failing (Dietz et al., 2016; Feng et al., 2024; Mandel et al., 2025).

3.2.3 Sources of loan failure

In Fig. 3 all the mechanisms whereby loans are defaulted on are show as outflows from the loan stocks. While all
355 nonperforming loans eventually default—except for those that may be reclassified as performing as a result of improved
liquidity conditions—all other classes of loans (performing, safe, and exploratory) can also fail under specific conditions. In
addition to conventional nonperforming loans failing, there are four other causes of defaults: interest rate hikes can cause

failures across every loan class, safe loans can also fail because of SLR, or R&D-driven innovation, and exploratory loans can be made vulnerable and fail from GDP deceleration.

Sudden changes in interest rates (see Section 3.4.2 for interest rate formation) directly impact loan performance. Interest rate cuts can make a share of nonperforming loans become performing by alleviating liquidity constraints (Bhandari and Weiss, 1993; Minsky, 2008). This is captured in Fig. 3 as the flow between performing and nonperforming loans. On the other hand, interest hikes increase debt servicing costs and can lead to failures across all loan classes, including safe loans, as they become too expensive to sustain (Adrian and Shin, 2008).

Higher sea levels increase the severity of floods, creating losses and damage in coastal areas in the absence of adaptation (see Ramme et al. (2025) for documentation of FRIDA's Sea Level Rise Impacts and Adaptation module). These losses and damages can strain even otherwise sound businesses, preventing them from servicing their loans and causing loan failures from the bank's perspective.

R&D-driven innovation, which renders a portion of existing investments obsolete (Diamond, 2006; Schumpeter, 1934, 1942), will cause further loan failure. Fig. 3 shows how the model captures this process with the outflow from safe loans. The value of the total number of safe loans issued by firms reflects the value of their productive assets. A portion of these assets is allocated to exploratory uses (R&D). The extent to which firms allocate these assets to R&D depends on their innovation orientation, which is determined in the Innovation submodule (Section 3.3). While R&D activities improve economy-wide productivity—as described in Section 3.3—the obsolescence it creates causes a share of safe loans to fail over time as new technologies diffuse and gradually displace previous techniques (Chinloy et al., 2020).

Finally, while exploratory loans can also fail from interest hikes, they mainly fail because of GDP deceleration. Fig. 3 shows how exploratory loan failure depends on GDP growth rate. This reflects the effect of liquidity constraints on high-risk innovative ventures (Brown et al., 2009; García-Quevedo et al., 2018).

3.3 Innovation and productivity

Following our Schumpeterian approach, we identify innovative investments from both incumbents and new entrants as a primary driver of long-term economic growth (Wong et al., 2005), structural change (Quatraro, 2016), and a disruptive process—reflecting creative destruction (Schumpeter, 1942). While the endogenous innovation process is spread across the various submodules, it is presented here under a single heading. This section describes the two classes of innovative investments and their disruptive effects, how they translate into potential productivity growth, and how climate impacts affect realised labour productivity.

Innovative investments in new market entrants are funded through exploratory lending as shown in Fig. 3. Such lending generates new investment opportunities, particularly in periods of high investment activity relative to growth potential—i.e., when the growth rate of nominal investment exceeds real GDP growth, exploratory lending increases (Gompers et al., 2008; Mendi, 2024; Nanda and Rhodes-Kropf, 2013). This reflects situations where credit expansion is not matched by real economic output growth, prompting the funding of innovative ventures that might open new avenues for additional profit. Conversely, a decline in the growth rate of annual bank profits places downwards pressure on exploration, as banks become less willing to take on more risk (Ahmad, 2021). Thus, exploratory loans fluctuate based on both financial and macroeconomic conditions.

Innovative investment by incumbents constitutes R&D activities and is represented as the reallocation of existing private assets, with implications for defaults in the finance module as described in Section 3.2.3. The driver of R&D activities and the resulting asset reallocations is located in their own Innovation submodule (not shown in the figures). Therein, the firm's innovation orientation increases (decreases) as a reaction to low (high) profit rates, thereby fluctuating with the business cycle (López-García et al., 2013; Mendi, 2024). The firms' orientation towards innovation is represented as an index. When the growth in firms' cash reserves slows down, firms rapidly become more inclined to innovate, reflecting an increased urgency to explore new potential revenue streams. Conversely, as cash reserves grow, this orientation declines gradually, indicating a preference for stability and risk aversion over uncertain R&D investments.

Productivity growth in the model arises from these innovative activities—exploratory loans and firm R&D activities (Hasan and Tucci, 2010; Kortum and Lerner, 1998)—and is simulated in the Employment submodule visualised in Fig. 5. The values of both exploratory loans and safe loans failing due to innovation are expressed as shares of GDP to enable direct comparison. On the basis of these normalised values, productivity growth is calculated, weighting the contribution of

exploratory loans more heavily to reflect their higher transformational potential (Acemoglu and Cao, 2015). After a time delay, realised productivity materialises, reflecting the time required for innovation to diffuse and yield economic benefits (Meade and Islam, 2006).

410 Finally, the realised labour productivity is also determined by climate impacts. Fig. 5 shows how realised labour productivity is influenced by both coastal flooding and rising temperatures. As the STA increases, labour productivity declines on the basis of the degree of exposure (Dasgupta et al., 2021). High-exposure labour, such as strenuous activity performed in the open, is the most affected. Low-exposure labour, such as work in not climate-controlled but shaded areas, is less affected. No exposure labour, such as work in a climate-controlled office setting, is not affected. To determine the allocation of global labour to these exposure classes over time, we model the development of the three main economic
415 sectors: agriculture, industry and services. For each of the sectors, we specify the share of high exposure, low exposure, and no exposure work and, thereby, the impact of climate change on labour productivity in each sector. Agriculture is the most susceptible to climate impacts, followed by industry, while workers in the service sector are less affected. The overall impact on labour productivity is then the average across sectors weighted by their share of the global economy. These shares are a function of GDP, reflecting the transition from agriculture to industry and services observed with higher economic output
420 (Alvarez-Cuadrado and Poschke, 2011).

3.4 Government

The Government submodule simulates fiscal and monetary policy, modelling the world's governments and central banks as aggregate agents responsible for managing tax revenue allocation, government debt dynamics, and the interest rate. Tax revenues collected from households and industry are redistributed through public investment, consumption, and transfers.
425 Total government expenditure levels adjusted in response to the public debt-to-GDP ratio.

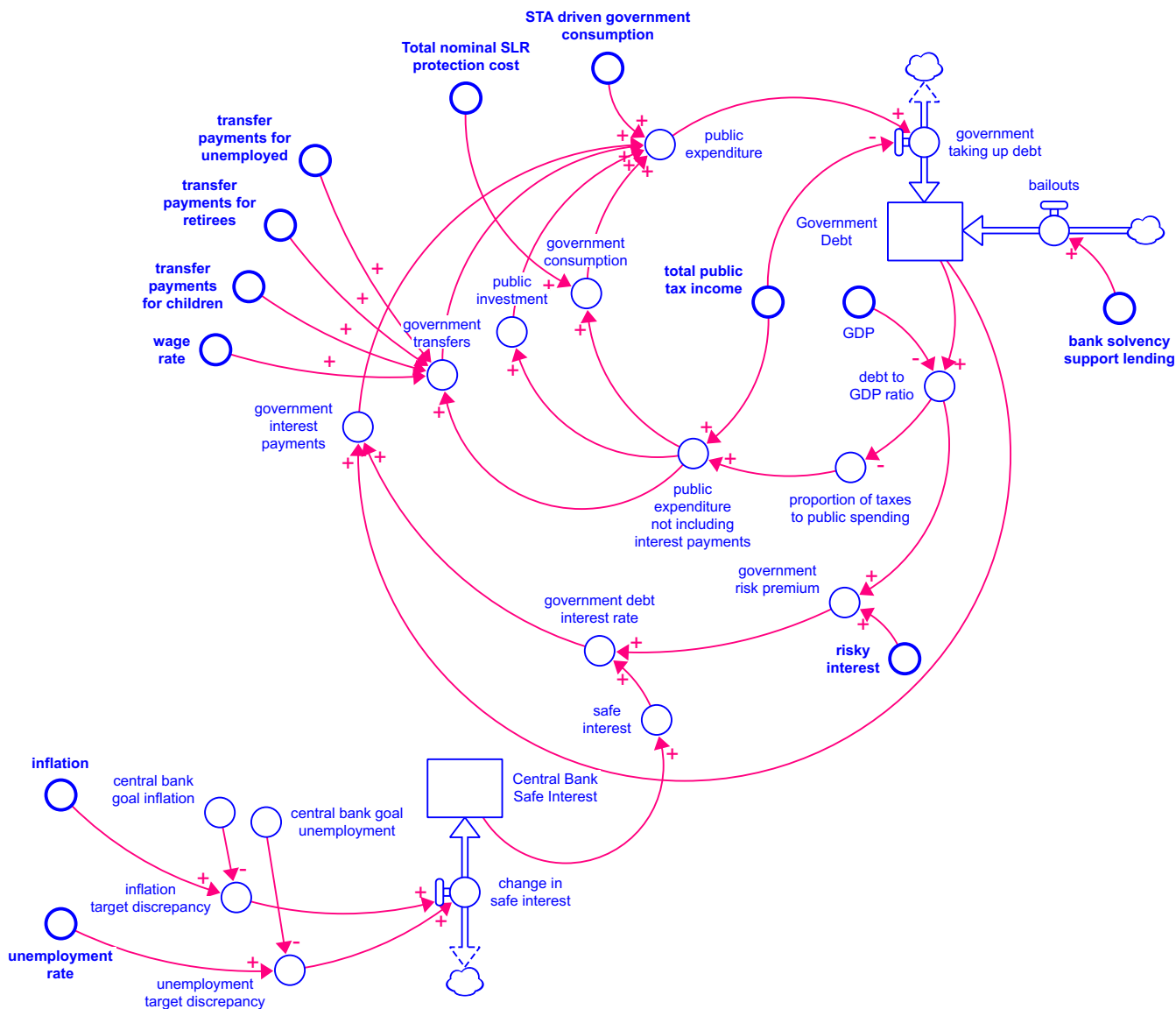


Figure 4: A simplified stock and flow diagram of the Government submodule. Bold elements receive inputs from elsewhere in the model.

3.4.1 Government revenue, expenditure and debt

Government expenditure, excluding interest payments on outstanding debt, shown as “public expenditure not including interest payments” in Fig. 4, is determined as a percentage of total tax income, which is levied on the flows of wages, profits, and rents (Narayan and Narayan, 2006; Ram, 1988). The percentage of tax income spent adjusts dynamically on the basis of the governments’ fiscal stance, measured by the debt-to-GDP ratio, which Fig. 4 shows to depend on the government debt stock and GDP. This ratio imposes constraints on the government’s expenditure relative to its tax-derived income: when the ratio is less than 1, spending exceeds tax revenue. However, as the ratio increases above 1, expenditure gradually decreases, eventually reaching a lower bound as the debt burden increases (Bohn, 1998; Ghosh et al., 2013).

The government is able to spend more than the sum of its income. The resulting deficit leads to the issuance of public debt from the government sector to the banking sector (Fang et al., 2025). It is important to note that we do not include nonbank holdings of government debt since the current version of the model lacks aggregated wealth portfolio management dynamics. Fig. 4 shows how government expenditures are modelled as the sum of investment, consumption, interest payments and transfer payments. In addition to deficits created via these regular spending and taxation processes, we model a second process for creating government debt, which is the additional government debt directly issued to stabilise the financial system should the banks be insolvent, effectively “bailing them out”. This is captured in Fig. 4 as the inflow “bailouts” to the government debt stock. Interest on outstanding government debt, including a risk premium (see Section 3.4.2 for interest rate formation), is assumed to always be paid, regardless of fiscal constraints. While this assumption ignores the empirical reality of defaulting sovereigns (Beers and Mavalwalla, 2017), the implementation of the latter is incompatible with the globally aggregate nature of the model.

After interest payments are accounted for, Fig 4. shows how government expenditure is allocated across investment, consumption, and transfers. The proportion of government expenditures directed to transfers is endogenised and adjusts dynamically with demographic changes and unemployment. FRIDA’s Demographics module (see Schoenberg et al. (2025b) for documentation) divides the global population into age cohorts, allowing the tracking of children and retirees to determine the cost of transfers to those cohorts. Welfare for the unemployed considers the number of unemployed individuals determined in the Employment module described in Section 3.5. The cost of the transfer payment per individual and each type of transfer—child support, pensions, and unemployment welfare—is adjusted with changes in the average wage rate with a delay (OECD, 2025). This ensures that transfers are adjusted for inflation in line with wages. Being partly driven by changes in unemployment, transfers are countercyclical (Chrysanthakopoulos et al., 2025). The remaining budgeted funds after transfers are then divided between investment and consumption. The allocation is dynamically adjusted in response to increases in STA; as climate-related losses and damage increase, consumption becomes a larger share of total expenditures to accommodate increased climate-driven repairs and maintenance (Qiao et al., 2015).

3.4.2 Central banking, the policy rate, and risk premia

The central banks are modelled as a single, aggregate agent and are tasked with keeping inflation and unemployment near their respective targets by adjusting the policy rate. This process is captured in Fig. 4 by the “Central Bank Safe Interest” stock which depends on measures of inflation and unemployment. Specifically, they aim for 2% inflation and 5% unemployment (Solow and Taylor, 1998). When inflation exceeds 2%, the policy rate is pushed upwards to slow growth in the price level, whereas inflation below 2% will push the rate down to stimulate investment (Clarida et al., 1998). Unemployment follows the same logic, but in reverse: if unemployment falls below 5%, it will increase the rate to mitigate excessive wage growth, whereas unemployment above 5% will decrease the rate to stimulate labour demand. These adjustments to the policy rate occur gradually, with greater weight given to inflation than unemployment (Cukierman and Lippi, 1999).

The private-sector average safe interest rate is derived as a moving average of the policy rate and shown in Fig. 3 as it is modelled in the Finance submodule. The average private sector’s risky interest is set above the safe rate by adding a risk premium to it. This premium changes dynamically to reflect the banks’ perceived risk profile, which is proxied by the default rate of loans relative to the total number of risky loans. As defaults increase in the Finance submodule (Section 3.2.3), the risk premium increases, making new loans more expensive (Huljak et al., 2022). Conversely, lower default rates reduce the risk premium, easing borrowing conditions (Fishman et al., 2024).

The interest rate on government debt, modelled in the Government submodule and show in Fig. 4, is similarly derived from a moving average of the policy rate, reflecting the composite nature of government debt duration (Dembiermont et al., 2015). It also includes a dynamically adjusted with a dynamic risk premium. When the debt-to-GDP ratio is at or below one, no extra risk premium is charged. When the ratio exceeds one, government debt is perceived to be riskier, and additional interest is charged in proportion to how much the ratio surpasses unity (Ardagna et al., 2007; Jacobs et al., 2020).

3.5 Employment

The Employment submodule simulates how the global working-age population transitions between nonactive, unemployed, and employed and how wage rates respond to changing labour market conditions. Also included within this submodule is our representation of labour productivity. Dynamics in this submodule are driven by public and private investment flows—originating in the Government (Section 3.4) and Finance (Section 3.2) submodules, respectively, as well as inflation, from the Inflation submodule (Section 3.6.). This submodule also integrates productivity gains related to “creative destruction”, where innovation from the Finance and Innovation submodules (Section 3.2. and 3.3, respectively) render existing processes obsolete but boost economy-wide productivity, with implications for worker displacement. This section details how these forces—labour supply and demand, wage formation, innovation and productivity—together shape global employment outcomes in the model.

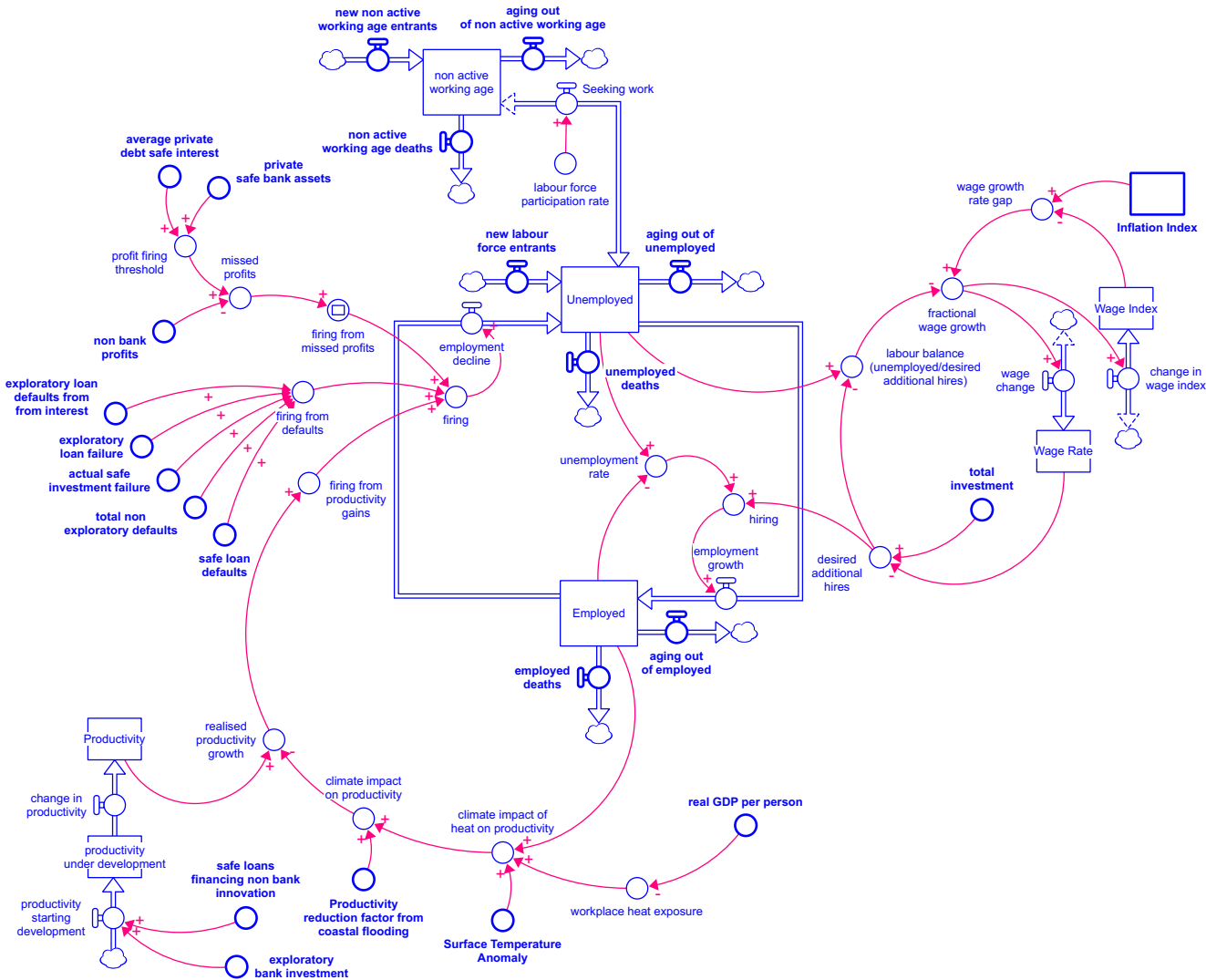


Figure 5: A simplified stock and flow diagram of the Employment submodule. Bold elements receive input from elsewhere in the model.

3.5.1 Labour and the wage rate

The working-age population, as represented in the Employment submodule, consists of people between the ages of 15 and 65. People ageing in or out or dying are modelled in the Demographics module (see Schoenberg et al. (2025b) for documentation). The working-age population is divided into three stocks shown in Fig. 5: nonactive, unemployed, and employed. The nonactive represents those of working age who cannot, or will not, seek work. Historically, approximately 35% of the working-age population has been in this category (ILO, 2024b). Hence, the model maintains a share of the working-age population in this category, with some allowance for variation during calibration given data uncertainties. The unemployed and the employed make up the labour force. Hiring and firing depend on the labour demand and wage dynamics.

Firms' labour demand stems from public and private investment, as determined in the Government and Finance submodules, respectively, calibrated as a fraction of total investment divided by the average wage rate (ILO, 2024). Fig. 5 shows how "desired additional hires" depends on total investment and the wage rate. The time required to hire new employees depends on the unemployment rate; as it declines, it becomes increasingly difficult to match vacancies with qualified workers (Mortensen and Nagypál, 2007).

The average wage rate responds asymmetrically to changes in the ratio of labour supply to labour demand: if the ratio falls below one, wages rise, whereas if it rises above one, wages decline at a lower rate, reflecting the relative "stickiness" of wages (Ehrlich and Montes, 2024; Grigsby et al., 2021). Fig. 5 shows how inflation also can influence the wage rate: if inflation (Section 3.6) outpaces nominal wage growth, the resulting purchasing power loss exerts upwards pressure on wages (Kahn, 1984). However, firms adjust wages with a lag to reflect real-world negotiation delays and the impact of existing contracts (Grigsby et al., 2021).

Firing moves workers from the employed to the unemployed category through the flow "employment decline" between the employed and unemployed cohorts in Fig. 5. Firing occurs for three reasons. The first reason is that defaults in the Finance submodule (Section 3.2.3) lead to layoffs proportional to the defaulted amount relative to the cost of labour. The second reason is that missed profits drop below a threshold tied to the average rate of interest on private debt (Section 3.4.2). The missed profits prompt firms to compensate by reducing their payroll (Coucke et al., 2007). The third reason is that productivity growth (Section 3.3) displaces portions of the employed over time, with delays reflecting the time it takes for new technologies and techniques to diffuse and result in firings (Feldmann, 2013; Quatraro, 2016).

3.6 GDP and inflation

GDP in this model is calculated in its own submodule (not shown in the figures) using a nominal expenditure approach, summing the final purchase price of public and private consumption and investment. Owing to the global nature of the model, international trade cancels out and is therefore not included. Private and public investments originate in the Finance and Government submodules, respectively (Sections 3.2 and 4.4). Private and public consumption originate in the Circular Flow (Section 3.1) and Government (Section 3.4) submodules, respectively. To derive real GDP in constant 2021 dollars, the model applies an inflation index calculated in the Inflation submodule described in detail below, ensuring comparability across simulation runs.

Inflation is calculated on the basis of two components: excessive income growth and input shocks (Auer et al., 2019; Deniz et al., 2016; Lim and Sek, 2015). Both components can be positive or negative, but inflationary pressures have an outsized effect compared with deflationary pressures, reflecting stickiness in price levels (Altonji and Devereux, 1999). Income growth leads to inflation when the combination of private income (from the Circular Flow submodule in Section 3.1.) and annual government deficit spending (from the Government submodule Section 3.4.) grows faster than economic expansion potential, proxied by productivity and employment growth, referring to qualitative development and quantitative growth, respectively (Hasan and Tucci, 2010). This imbalance places upwards pressure on the price level. The input-shock components of inflation originate outside the core Economy module—specifically from the Energy and Land Use and Agriculture modules (see Schoenberg et al. (2025b) for documentation). When demand exceeds supply in animal products and crops, it adds to inflation. The opposite has a deflationary effect. Other pressures arise from the growth rates of fertiliser use, cropland, and grazing land and the marginal cost of energy. Changes in these growth rates exacerbate scarcity, driving

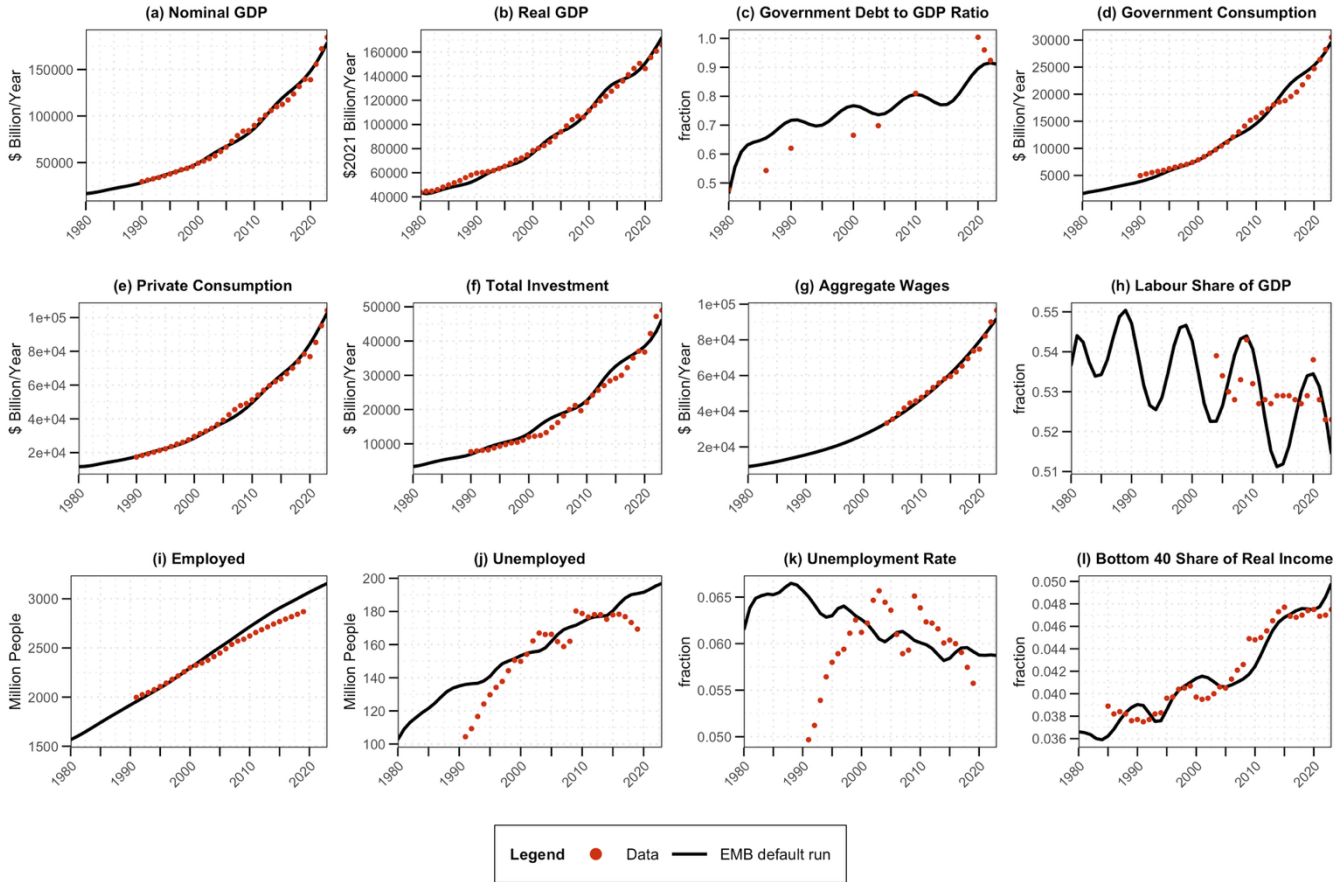
up input costs and triggering cost-push inflation. Furthermore, they proxy all historical input-shock inflation, and their inflationary contribution is weighted accordingly, capturing the impact of scarcity on real growth (Parker, 2017).

545 **4 Integration, calibration and uncertainty**

FRIDA is implemented in Stella Architect 3.8 and simulated over a time horizon spanning 1980–2150. The integration method used to simulate the model is fourth-order Runge-Kutta (RK4) with a timestep of 1/8th of a year. A multistep protocol was developed for estimating parameter value ranges and presenting results in such a way as to communicate the range of possible outcomes given the input uncertainty rather than a single outcome (Schoenberg et al., 2025b).

550 The first step of the protocol is calibration using Powell’s BOBYQA algorithm (Powell, 2009). Parameter inputs whose values, or value uncertainty ranges, could not be sourced from the data or literature were given wide uncertainty ranges on the basis of the outcomes created at the most extreme potential values for the parameter. The calibration involved varying the inputs within their individual uncertainty ranges to minimise the squared error between historical data and simulated data for 16 payoff elements for the period 1980–2023. The temporal coverage of the historical data series varies.

555 The labour share of GDP for instance, sourced from the ILO, begins in 2004, which accounts for the absence of observed data before that year in Fig. 6g and h. A table listing all the payoff elements, with reference to the corresponding datasets, is found in Appendix A. This first step produced the best fit single-term endogenous model behaviour (EMB) shown in Fig. 6 for a selection of 12 key performance indicators.



560 **Figure 6: Best fit FRIDA results for key macroeconomic indicators against observed data.**

The single-run EMB run, as shown in Fig. 6, presents the model's behaviour over the 1980–2023 calibration period, with simulated trajectories whose general level and variability are broadly consistent with the observed data. Notable features of these results include the model's ability to produce business cycle fluctuations, which are visible in the changing slope of GDP (Fig. 6a and 6b) and are particularly prominent in the unemployment rate (Fig. 6k) and the labour share of GDP (Fig. 6h). These cycles arise from the interaction of innovation and finance described in Sections 3.2 and 3.3, and should not be interpreted as replicating specific historical downturns. Their amplitude and frequency are broadly consistent with observed macroeconomic variability, but their phase does not align perfectly with particular historical economic events, since FRIDA does not include shocks exogenous to the processes described in Section 3. This explains the absence of effects of the 2020 pandemic, 2009 banking crisis and similar events in the model results. For the historical period, the model does not include discrete policy changes, and endogenous policy changes are continuous. These deviations are consistent with FRIDA's design philosophy: rather than aiming for perfect replication of historical events driven by exogenous shocks, the model generates endogenous oscillations through the feedback processes within and across economic sectors.

The second step of the protocol involves exploring the possibility space, arising from the parameter uncertainty ranges. This is done in two steps. In the first step, the parameters are constrained to their likely ranges. The likely range is defined as the interval within which a parameter can be varied, while the likelihood of the model's output remaining above the 1/1000th of the best fit likelihood obtained during calibration. These ranges are then symmetrised within the originally provided widest possible bounds. This approach ensured that samples were distributed evenly around the default parameter set while still covering the regions with the highest likelihood. The resulting parameter bounds are documented in Appendix B.

On the basis of these ranges, we conduct a 100,000-member global sensitivity analysis (Saltelli, 2008) using Sobol sequence sampling (Sobol' and Levitan, 1999). This technique produces a set of sample points aimed at efficiently sampling the high-dimensional parameter space of the model. This method is needed because the high number of parameters to be sampled makes more systematic sampling computationally infeasible. The model is then run for each sample point, i.e., a set of parameter values, to produce the ensemble presented as the result of the model. As a result, the ensemble runs are not weighted by their likelihood, yet they can still be understood probabilistically: the resulting bounds highlight the regions of the output space that the ensemble most frequently explored, providing a probability-based interpretation conditioned on our sampling approach. It should be noted, however, that these confidence bounds are wider than they would be if likelihood weighting had been applied. While the sample is not weighted by likelihood, the spread captures a wide range of plausible system behaviours. The ensemble median, along with the 67% and 95% confidence intervals, are reported in Section 5.

This protocol reflects the project's commitment to openly communicating uncertainty, and it also underscores FRIDA's nature as an exploratory model: the goal is not to provide point predictions but to test the robustness of insights across a meaningful uncertainty space. Presenting single-line projections would understate the range of potential futures and could mislead stakeholders about the model's precision.

5 Simulation results

This section evaluates the model behaviour of the FRIDA v2.1 Economy module over a simulation time horizon spanning 2020–2150. The results are based on the 100,000-member ensemble run derived from the simulation protocol described in Section 4. Fig. 7 shows the distribution of the ensemble members results for real GDP in selected years. As the runs progress the spread increases, exhibiting a fat tail. The ensemble median remains close to the calibrated best fit throughout the simulation period. This is a result of the uncertainty procedure varying the parameters symmetrically around the calibrated values. In the following figures the best fit run is not explicitly shown. It is treated as merely one of the ensemble members.

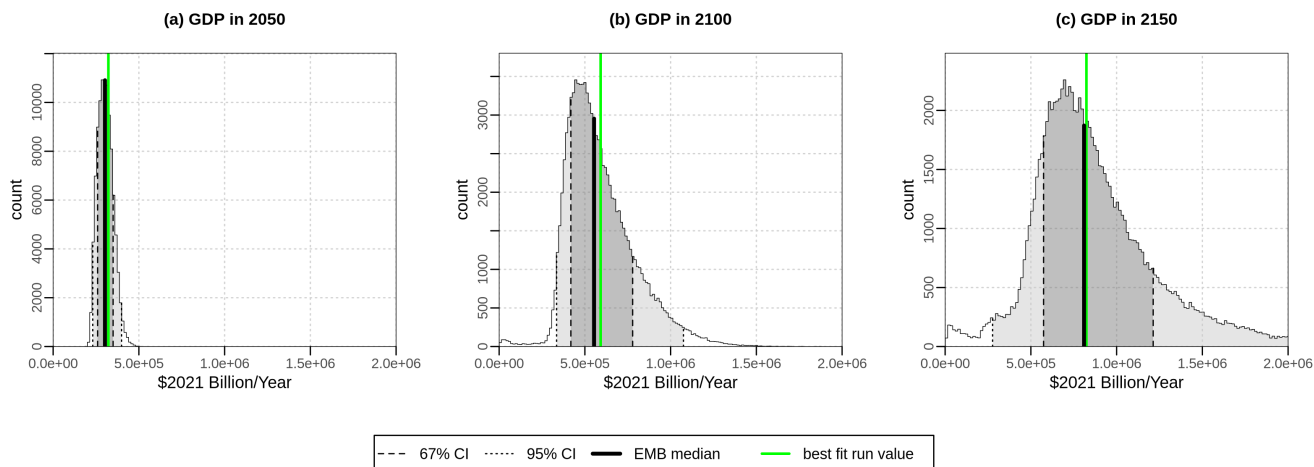


Figure 7: Histogram of the distribution of real GDP in 2021\$ across the 100,000 ensemble members in the years 2050 (a), 2100 (b), and 2150 (c). The black solid line shows the median, darker shaded areas show the 67% confidence intervals, and lighter shaded areas show the 95% confidence intervals. The green line indicates the value of the calibrated best fit estimate in these years.

Figs. 8 to 10 present the baseline results of the 100,000-member ensemble runs for key macroeconomic indicators. While Fig. 6 above presents FRIDA's behaviour over the historical calibration period, here we show the projected future economic trajectories under uncertainty. The figures show the medians and 67% and 95% confidence intervals of the 100,000 ensemble projections. While individual ensemble members have the oscillatory behaviour discussed above, the figures, especially Fig. 8, display relatively smooth trajectories as they show statistical measures of the asynchronous oscillatory behaviour across ensemble members. The smoothness is therefore a property of the ensemble statistics rather than of the model dynamics. Each member oscillates, but members differ in the period, amplitude, and phase of their cycles because they are generated under different parameterisations. When the median is taken across members at each point in time, peaks in some members coincide with troughs in others, flattening it. The same characteristic widens the confidence intervals, since the spread reflects both projection uncertainty and the different cycle phases across the ensemble.

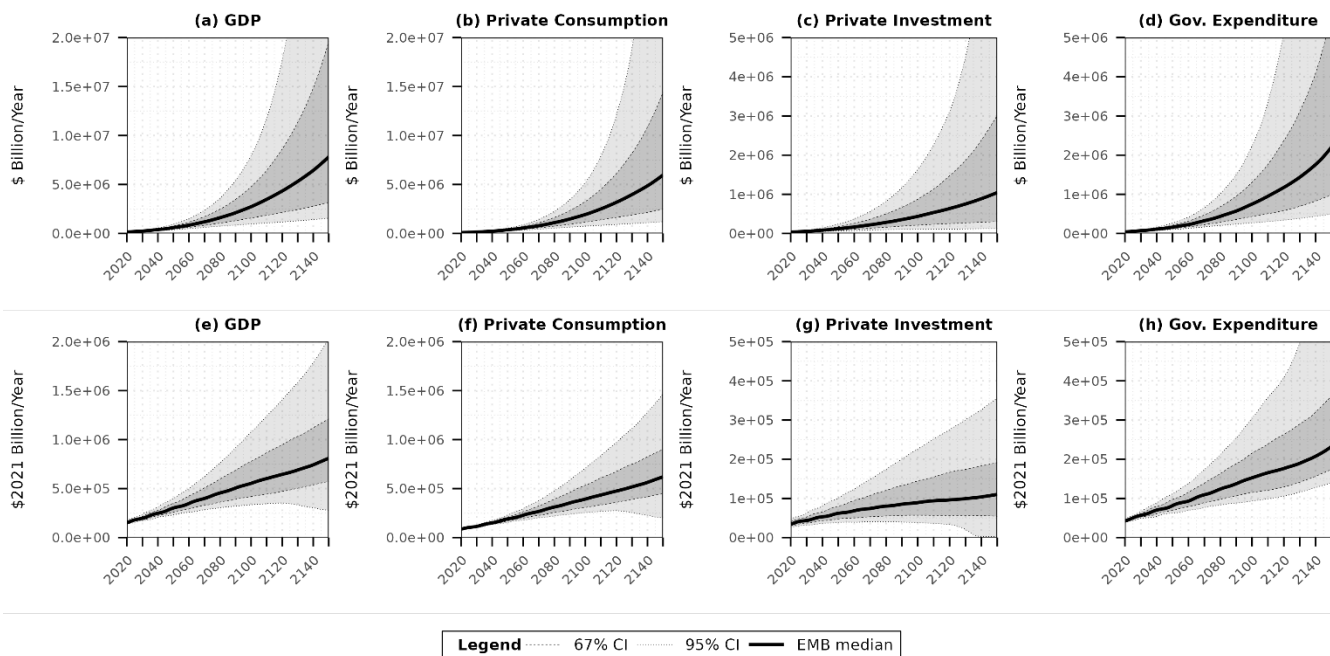
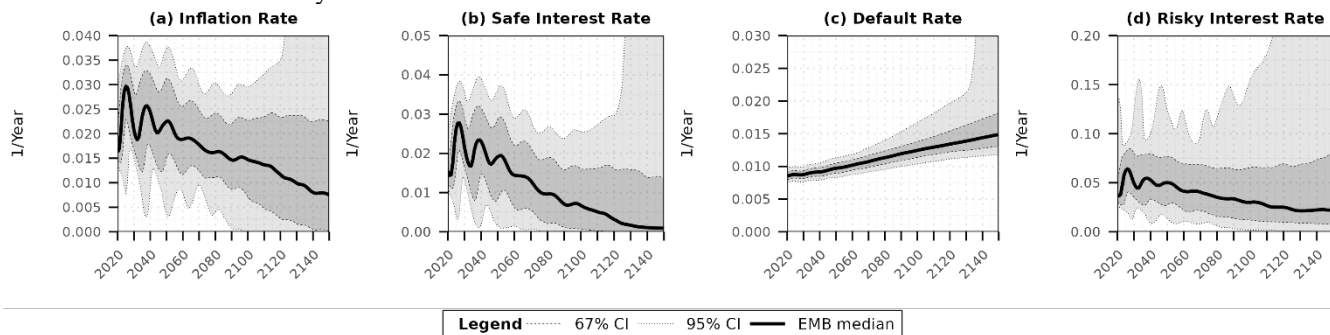


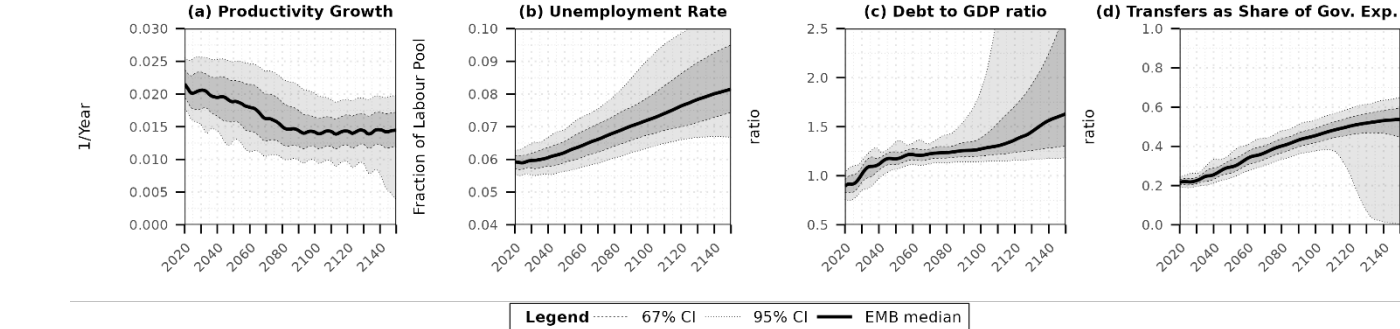
Figure 8: Nominal and real GDP and their components across the 100,000-member ensemble. Solid lines show the median, darker shaded areas show the 67% confidence intervals, and lighter shaded areas show the 95% confidence intervals.

Fig. 8 shows GDP and its components in both nominal and real terms (constant 2021\$). In nominal terms, private consumption (Fig. 8b), private investment (Fig. 8c), government expenditure (Fig. 8d), and, hence, GDP (Fig. 8a) increase exponentially, which is consistent with historical patterns. Much of this growth, however, is due to inflation (Fig. 8a). Nonetheless, most ensemble members also exhibit continuing growth in real GDP (Fig. 8e) until the end of the simulation time horizon, albeit not exponentially. Private consumption (Fig. 8f) is the largest component of GDP and the main driver of the long-term upwards trend in GDP. The decline in the median growth rate around the end of the century is due to a decline in the growth of private consumption and investment (Fig. 8f and 8g). The investment slowdown originates in the Finance submodule (Section 3.2), where the rising STA increases the failure rate on new loans. The result is higher lending standards and risk premia, which together dampen private investment and income growth. This slowdown is partly compensated by an increase in government expenditures, including public investments (Fig. 8h). This is in turn driven by increases in government transfers, described below. However, there are ensemble members within the 95% confidence interval that exhibit negative growth in real terms (nominal growth remains positive), reflecting ensemble members whose parameterisations specify the strongest climate damage consistent with historical data when combined with uncertainty in other areas of the economic system.



635 **Figure 9: Key variables of the Finance submodule across the 100,000-member ensemble. Solid lines show the median, darker shaded areas show the 67% confidence intervals, and lighter shaded areas show the 95% confidence intervals.**

The inflation rate (Fig. 9a), starting from historical levels, declines over time. As a consequence, the central bank safe interest (Fig. 9b) rate is reduced (see Section 3.4.2). Without quantitative easing measures (not currently implemented in FRIDA), the central bank is not able to maintain the inflation rate at the target of 2%, with the safe interest rate near the zero lower bound towards the end of the simulation time horizon. Interest rates paid by private investors (Fig. 9d) do not decline as strongly as the safe interest rate does because of the risk of defaults (Fig. 9c) perceived by banks, driving a risk premium between safe and risky interest rates (see Section 3.2). The default rate itself is climate-sensitive: rising STA increase defaults on new loans, and the resulting defaults sustain the risk premium even as the safe interest rate falls. The widening gap between the safe and risky rates is therefore a direct channel through which climate impacts are transmitted to the cost of investment finance.



650 **Figure 10: Changes in labour productivity, unemployment rate, government debt to GDP, and the share of government expenditure spent on transfers across the 100,000-member ensemble. Solid lines show the median, darker shaded areas show the 67% confidence intervals, and lighter shaded areas show the 95% confidence intervals.**

A share of private investment (Fig. 8c) flows into research and development with the potential to increase productivity. Fig. 10a shows the effect of these exploratory loans on changing economy-wide productivity (see Section 3.3). Increases in productivity increase GDP but also contribute to redundancies among workers, contributing to an increasing unemployment rate. The unemployment rate is further affected by demographic contraction around the turn of the century, combined with downwardly sticky wages keeping labour costs relatively high and decreasing the investment growth rate, leading to fewer new job vacancies because of proximately default-driven, but ultimately climate-driven, economic slowdown (see Section 3.2.3). Furthermore, economic volatility causing defaults translates to increased firing due to missed profits, in addition to productivity-driven displacement. These factors combine to reduce the economy's capacity to finance ever-growing numbers of workers, as clearly reflected in the growing unemployment rate (Fig. 10b; see Section 3.5). The increasing unemployment together with demographic change (see Schoenberg et al. (2025b)) drives an increase in government expenditure for transfer payments (Fig. 10d). Transfers are designed to be countercyclical in the model: the share of expenditure directed to transfers adjusts with the number of unemployed and with the minors and retiree cohorts, so a rising unemployment rate and a rapidly ageing population both enlarge public spending. Increasing transfer payments contribute to the government debt-to-GDP ratio (Fig. 9c). Increasing government debt increases government interest rates, further straining the budget, which reduces the government's ability to invest and pay for transfers and other government services.

Taken together, these results show that the FRIDA v2.1 Economy module produces internally consistent long-run macroeconomic dynamics under uncertainty. The widening confidence intervals for the results reflect the substantial uncertainty inherent in long-run economic projection under climate constraints. However, robust patterns do emerge: median nominal aggregates continue to rise, and real activity generally increases but slows towards the end of the century because financial, demographic, and labour-market constraints are sensitive to climate impacts. While median projections suggest a generally growing but decelerating economy, the ensemble intervals expose substantial risk arising from climate impacts on

the financial system, which can generate persistent unemployment, increase transfer burdens and public debt, and halt growth.

675 6 Conclusion

A substantial body of research has identified critical limitations in contemporary IAMs: aggregate damage functions that operate as statistical black boxes, the absence of financial system representation despite mounting evidence of climate-related financial risks, a lack of business cycle dynamics and short-term mechanisms with long-term consequences, and exogenous treatment of innovation. The novel economy module of FRIDA v2.1 documented in this paper addresses these
680 limitations. The module is defined by its institutionally disaggregated structure, which allows for the traceability of specific climate impact and policy pathways through distinct economic sectors and processes. Moreover, it enables the model to leverage more observable economic variables with available time series data, strengthening its empirical foundations beyond the highly uncertain parametrisations of aggregate damage functions.

While FRIDA v2.1 addresses several critical gaps in the IAM literature, it introduces others that are inherent to its
685 approach. Its global aggregation masks local climate impacts, precluding geographically differentiated policy analysis. The disequilibrium framework presents two distinct challenges. First, historical calibration necessarily attempts to reproduce observed fluctuations—including exogenous shocks such as pandemics and wars outside the model's scope—using only endogenous mechanisms, potentially overfitting to historical volatility. The resulting calibration should therefore not be interpreted as a replication of specific historical episodes, and the endogenous oscillations the model generates do not, and
690 are not intended to, align perfectly in phase with observed business cycles. Second, the absence of utility maximisation assumptions limits the identification of optimal policy sets. Our ensemble methodology partially addresses these limitations by presenting results as probability distributions rather than point forecasts, explicitly foregrounding the considerable uncertainties inherent to long-run economic projection.

The model's primary contributions lie in its endogenous treatment of economic processes and the inclusion of
695 finance, innovation, and business cycles. FRIDA v2.1 adopts a Schumpeterian disequilibrium framework that explicitly represents the role of the financial sector in both short- and long-term dynamics. This enables the analysis of climate-related financial risks, transition funding, and the potential for disorderly transitions to trigger system-wide instability—dynamics that remain largely unexplored in current policy-relevant models. The endogenous representation of innovation, driven by both incumbent R&D activities and the financing of new market entrants, captures the inherently disruptive nature of
700 technological change and its dual role in both promoting productivity growth and displacing existing investments and workers.

The simulation results show that the model produces dynamics broadly consistent with observed macroeconomic patterns over the calibration period 1980–2023, and generates plausible projections through 2150. The ensemble approach, which is based on 100,000 simulation runs, produces widening confidence intervals over time, reflecting the compounding
705 uncertainties as economic systems encroach on planetary boundaries. The median projections show that nominal aggregates increase throughout the simulation, whereas real activity expands more slowly and begins to decelerate as financial, demographic, and climate pressures accumulate. Moreover, the ensemble's more extreme scenarios reveal a substantial risk of economic volatility and even decline under more adverse climate impacts.

Following the historical period, the ensemble also reveals long-run constraints that arise from the model sectors' endogenous connections. Rising STA gradually weakens firms' ability to sustain investment by increasing failure rates and slowing growth. Moreover, demographic ageing and labour-market pressure increase the share of government spending devoted to transfers. Governments can rely on debt to maintain spending, but doing so increases interest burdens and inflationary pressure, further tightening fiscal conditions. These patterns—deteriorating investment conditions, rising welfare costs, and shrinking discretionary expenditures—emerge directly from the model's feedback structure rather than from
715 imposed assumptions.

These endogenous constraints offer unique realism for policy experimentation with FRIDA v2.1. Climate-transition policy scenarios will operate within and interact with evolving financial, fiscal, and demographic pressures rather than being assessed in isolation. As a result, scenario outcomes reflect genuine trade-offs—for example, between maintaining welfare commitments, funding mitigation and adaptation, and preserving financial stability. By allowing these tensions to arise from

720 the system’s internal dynamics, FRIDA v2.1’s economy module offers a rich exploration of transition pathways by embedding policy choices within the evolving macro conditions that will ultimately govern their success.

Appendix A: Payoff elements for calibration

Table A1: Calibration payoff elements, with dataset citations and notes

Variable name in source code	Source	Notes
Circular flow submodule		
private_consumption	(World Bank, 2023b, d)	Calculated
Labour_share_of_GDP	(ILO, 2024c)	
bottom_40_share_of_real_income	(WID, 2024)	
Finance submodule		
total_investment	(World Bank, 2023b, c)	Calculated
government_consumption		
debt_to_gdp_ratio	(Poplawski-Ribeiro et al., 2023)	
Employment submodule		
Employed	(ILO, 2024a)	
Unemployed	(ILO, 2024d)	
Unemployment_rate	(ILO, 2024a, d)	Calculated
annual_aggregate_wages	(ILO, 2024c; World Bank, 2023b)	Calculated
GDP submodule		
nominal_GDP	(World Bank, 2023b)	
real_GDP_in_2021_c\$	(World Bank, 2023a)	
nominal_gdp_growth_rate_deviation_from_average_in_calibration_period_summed	(World Bank, 2023b)	Calculated
real_GDP_growth_rate_deviation_from_average_in_calibration_perdioid_summed	(World Bank, 2023a)	Calculated
Inflation submodule		
gdp_difference	n/a	real GDP base year

Table B1: Uncertain parameters with their estimated values and ranges rounded to three decimal places

Name in source code	Value	Min	Max
consumption_habits_adjustment_time	1.464	1.241	1.687
desired_savings_as_multiple_of_profits	12.006	11.989	12.023
desired_savings_as_multiple_of_wages	1.198	1.190	1.205
fraction_of_income_to_consumption	0.750	0.748	0.752
fraction_of_owner_income_to_consumption	0.750	0.747	0.753
initial_firms_checking_accounts	5003.534	4979.060	5028.009
initial_owner_savings	46357.047	46313.514	46400.579
initial_worker_savings	22209.036	22175.118	22242.955
normal_bottom_40_wages	0.116	0.114	0.118
profit_tax_rate	0.335	0.333	0.336
time_for_bottom_40_to_be_affected	2.717	1.328	4.106
time_for_inflation_to_change_consumption	1.000	0.981	1.019
time_to_consider_desired_savings	2.373	2.308	2.438
time_to_pay_out_profits	2.522	2.511	2.532
time_to_reach_savings_goal	4.981	4.961	5.000
wage_tax_rate	0.213	0.210	0.216
xmiddle	0.743	0.736	0.751
xspeed	14.664	13.704	15.624
yfrom	1.672	1.570	1.774
yto	0.579	0.567	0.591
average_time_private_refinances_debt	3.000	2.772	3.228
bank_desired_asset_to_liability_ratio	1.100	1.099	1.101
bank_reference_formation_time	80.000	66.465	93.535
effect_of_default_rate_on_lending_standards_curvature	4.004	3.912	4.096
effect_of_default_rate_on_lending_standards_range	1.947	1.923	1.972
effect_of_lending_standards_on_failure_rate_curvature	5.915	5.677	6.154
effect_of_lending_standards_on_failure_rate_range	2.950	2.836	3.064
elasticity_of_bankruptcy_rate_on_risk_premium	2.960	2.859	3.061
elasticity_of_desired_bank_growth_rate_to_changes_in_lending_standards	2.983	2.913	3.053
equity_gap_above_min_effect	0.434	0.376	0.492
equity_gap_below_max_effect	3.105	2.409	3.802
equity_gap_effect_max_x_magnitude	0.170	0.152	0.188

equity_ratio_gap_effect_above_exponent	-1.132	-1.265	-1.000
equity_ratio_gap_effect_below_exponent	1.071	1.000	1.142
exploration_effect_midpoint	1.211	1.205	1.217
exploratory_premium	2.000	2.000	2.000
exponential_effect_of_interest_on_defaults	9.049	8.718	9.380
failure_rate_floor	0.002	0.001	0.003
initial_annual_bank_negative_profit_fractional_growth	0.140	0.050	0.230
initial_bad_loans	202.031	200.716	203.346
initial_bank_investment_growth_rate	0.084	0.083	0.084
initial_fraction_of_safe_loans_financing_innovation	0.030	0.027	0.033
initial_fraction_of_safe_loans_that_are_exploratory	0.020	0.017	0.023
initial_good_loans	8240.068	8193.957	8286.179
initial_net_bank_asset_fractional_growth	0.155	0.154	0.156
initial_non_energy_bank_investment	3000.000	2989.476	3010.524
initial_risk_premium	0.028	0.027	0.030
Initial_safe_assets	72569.200	72508.643	72629.757
loan_maturations_time	11.561	9.000	13.000
max_change_in_failure_rate	0.076	0.073	0.079
max_change_in_lending_standards	0.227	0.224	0.230
max_risk_premium_increase_rate	0.015	0.015	0.015
normal_bank_profit_payment_fraction	1.000	0.993	1.007
normal_exploration_rate	0.037	0.037	0.037
normal_failure_rate	0.037	0.030	0.045
normal_growth_rate_of_GDP_growth_rate	1.000	0.826	1.174
normal_rate_of_converting_safe_loans_to_safe_loans_for_financing_innovation	0.030	0.029	0.031
renegotiation_time	1.000	1.000	1.000
reporting_delay	0.475	0.410	0.550
safe_asset_default_rate_threshold	0.015	0.015	0.015
scale_of_exploratory_loan_failure	7.508	6.016	9.000
sensitivity_of_bank_exploration_to_profit_decline	0.502	0.485	0.519
sensitivity_of_bank_profit_payment_to_net_asset_growth	1.100	1.093	1.107
sensitivity_of_defaults_to_changes_in_risky_interest	11.899	11.436	12.363
sensitivity_of_effect_of_safe_default_rate_on_lending_standards	0.151	0.146	0.156
sensitivity_of_effect_of_sta_on_failure_rate	0.567	0.350	0.850
sensitivity_of_effect_of_stranded_energy_capital_on_other_assets	2.156	1.000	3.313
sensitivity_of_exploratory_loan_failure_to_GDP_growth	1.049	1.008	1.089

sensitivity_of_failure_rate_to_investment_to_gdp_ratio	0.170	0.143	0.198
sensitivity_of_investment_and_GDP_difference	2.217	2.157	2.277
time_for_safe_loans_that_will_fail_to_actually_fail	13.752	13.527	13.977
time_to_change_lending_standards	1.140	1.119	1.162
tolerance_of_the_cap_on_indicated_risk_premium	0.010	0.005	0.015
average_time_government_refinances_debt	7.000	5.977	8.023
averaging_time_to_adjust_tax_based_on_income	2.284	2.219	2.349
baseline_interest	0.020	0.019	0.021
central_bank_adjustment_time	1.500	1.305	1.695
central_bank_interest_rate_reactivity	1.000	0.936	1.064
fraction_of_wages_as_child_transfer_payments	0.082	0.079	0.086
fraction_of_wages_as_pension	0.375	0.336	0.413
fraction_of_wages_as_unemployment_payments	0.250	0.176	0.324
inflation_weight	1.330	1.260	1.400
initial_government_debt	9712.900	8395.455	11030.345
initial_measured_GDP	20970.400	20000.000	21940.800
initial_safe_interest	0.039	0.038	0.039
initial_smoothed_total_public_tax_income	2635.200	2584.367	2686.033
maximum_effect_of_debt_to_GDP_ratio_on_government_spending	1.245	1.237	1.252
normal_share_of_public_expenditure_available_for_investment_and_consumption_to_consumption	0.835	0.828	0.842
sensitivity_of_banks_gov_debt_risk_perception	0.100	0.050	0.150
sensitivity_of_STA_on_public_consumption	0.071	0.054	0.087
time_for_government_to_change_transfers	5.000	3.123	6.877
time_to_measure_unemployment	1.000	0.905	1.095
unemployment_weight	0.330	0.295	0.365
average_development_completion_time	5.000	4.746	5.254
fraction_of_productivity_gains_translating_to_firing	0.500	0.466	0.534
hiring_to_investment_ratio	0.455	0.450	0.459
initial_productivity_growth	0.026	0.026	0.026
initial_unemployment_rate	0.061	0.060	0.063
Initial_Wage_Rate	5695.751	5680.618	5710.883
labour_force_participation_fixed	63.939	63.615	64.263
max_change_in_fractional_wage_growth	0.069	0.068	0.070
negotiation_effectiveness	1.000	0.643	1.357
productivity_yield_of_exploratory_investment	1.947	1.938	1.957
productivity_yield_of_non_bank_innovation	0.133	0.132	0.134

profit_threshold_multiplier	1.000	0.769	1.231
rent_to_investment_ratio	0.250	0.248	0.252
sensitivity_of_firing_to_profit_discrepancy	0.500	0.250	0.750
sensitivity_of_hiring_to_unemployment_rate	2.290	2.262	2.317
share_of_employment_in_agriculture_intercept	62.902	62.005	63.799
share_of_employment_in_industry_intercept	-2.581	-3.371	-1.791
share_of_employment_in_industry_slope	2.931	2.875	2.987
share_of_employment_in_services_intercept	-30.739	-31.508	-29.970
share_of_employment_in_services_slope	7.498	7.435	7.562
Temperature_effect_on_high_exposure_productivity	-6.133	-7.823	-4.442
Temperature_effect_on_low_exposure_productivity	-2.083	-4.666	0.501
Temperature_squared_effect_on_high_exposure_productivity	-1.062	-1.500	-0.624
Temperature_squared_effect_on_low_exposure_productivity	-1.569	-2.000	-1.139
threshold_unemployment_rate	0.099	0.099	0.100
time_for_defaults_to_affect_firing	1.000	0.500	1.500
time_seeking_employees	1.417	1.391	1.443
time_to_fire_from_missed_profits	1.000	0.500	1.500
Urgency_with_which_wage_growth_is_demanded	2.000	1.000	3.000
wage_adjustment_negotiation_time	2.000	1.000	2.999
initial_nominal_GDP_growth_rate	0.041	0.039	0.044
initial_real_GDP_growth_rate	0.024	0.007	0.041
Agriculture_share_of_GDP_intercept	3.449	3.017	3.880
agriculture_share_of_GDP_slope	-2.288	-2.570	-2.006
agriculture_share_of_GDP_starting_point_determinant	19.877	19.754	20.000
deflation_adjustment_time	3.825	3.803	3.848
inflation_adjustment_time	1.451	1.401	1.500
initial_animal_products_demand_growth_rate	0.020	0.005	0.035
initial_animal_products_production_growth_rate	0.035	0.020	0.050
initial_crop_demand_growth_rate	0.020	0.005	0.035
initial_crop_supply_growth_rate	0.020	0.005	0.035
initial_employed_growth	-0.015	-0.016	-0.014
initial_energy_demand_growth_rate	0.020	0.019	0.021
initial_energy_supply_growth_rate	0.020	0.019	0.021
initial_growth_rate_in_cropland	0.003	0.000	0.006
initial_growth_rate_in_grazing_land	0.001	0.000	0.002
initial_income_growth_rate	0.020	0.019	0.021

initial_inflation_index	0.383	0.382	0.384
initial_inflation_rate	0.061	0.061	0.062
time_to_measure_growth_rate	2.000	1.500	2.500
weight_of_cropland_inflation_contribution	0.094	0.066	0.123
weight_of_fertilizer_used_per_unit_of_crop_production	0.018	0.001	0.035
weight_of_grazing_land_inflation_contribution	0.054	0.001	0.108
weight_of_irrigation_water_used_per_unit_of_crop_production_inflation_contribution	0.050	0.036	0.064
weight_of_marginal_energy_cost_inflation	0.812	0.791	0.833
reference_cash_reserve_growth_rate	0.049	0.042	0.056
sensitivity_of_effect_of_cash_reserve_growth_rate_on_firms_innovation_orientation	-0.114	-0.156	-0.073
time_for_firms_to_adjust_innovation_orientation	3.005	2.000	4.009

730 Table C1: Comparison of selected climate--finance/SFC macroclimate models across key dimensions.

Model Name	Reference	Model Type	Finance	Inflation	Business Cycles	Innovation	Damages	Uncertainty	Time Horizon	Coupling of Earth System and Human System
FRIDA v2.1	This Paper	SD	Private investments at risky interest rates driven by loan standards responding to default rates. Central Bank managing safe interest rates targeting inflation and unemployment. Gov. Bailouts of failing banks.	Determined by supply demand imbalances and costs of energy and crop production.	Emergent property of delays in the perception of loan defaults and investment behaviour.	Driven by exploratory investments.	Loan maturities related to climate impacts. Labour productivity reduced by temperature exposure. Energy demand and energy efficiency cause	Large scale sampling of parametric uncertainty.	1980--2150	Fully coupled Fair climate module
DSK	Reissl 2025	SD-ABM	Agent based banks providing loans to investing firms. Central Bank managing safe interest rates targeting inflation and unemployment. Gov Bailouts of failing banks.	Based on agent based mark-up pricing	Emergent property of agent interactions: Innovation, Investment, Credit constraints, Market Selections, Heterogenous Agent Introductions	Capital goods firms endogenous investments in R&D	Labour productivity, energy efficiency, capital shocks, inventory shocks	Stochastic processes sampled through monte carlo runs	400 periods (quarters) > 2000--2100	Fully Coupled Very simplified climate (one box climate model)
DEFINE & Nikolaïdi 2022		SD	Commercial banks create credit endogenously, subject to credit rationing and capital constraints. Central bank provides liquidity. Firm defaults.	Not represented.	Emergent property of agent interactions.	Only for green capital	Aggregate damage function (DICE type)	Policy Scenarios, no uncertainty	2021--2100	Fully Coupled Very simplified climate (one box climate model)
N/A	Bovari et al. 2020	SD	Stock-flow consistent banking system with private debt, bank lending, credit rationing, and an endogenous short-term interest rate. Central bank setting the safe interest rate following a Taylor rule	Phillips-curve-style wage dynamics and price adjustment toward a markup over unit labor costs	Endogenous nonlinear dynamics from interactions among output, employment, wages, profits, investment, debt, and interest rates	exogenous: labor productivity grows at a constant rate, and clean technology	Climate damages reduce output and, in some scenarios, capital stock; modeled with Nordhaus and Diez-	Monte Carlo simulation over uncertain climate and productivity parameters such as climate sensitivity and carbon-cycle	From the initial calibration period to 2100, with focus on 2050 and 2100	Fully Coupled (two layer climate module)
DICE	Nordhaus 2019	Ramsey Growth Model	Investments chosen by Social Decision Maker (SDM), no finance	Not represented	Not represented.	Exogenous over time	Aggregate damage function	Parametric uncertainty in expected utility maximisation framework.	--2200	Fully Coupled, one box climate

- Code and data availability.* FRIDA is released as a free and open-source model on GitHub <https://github.com/metno/WorldTransFRIDA>. The specific version used for this manuscript is available on Zenodo <https://doi.org/10.5281/zenodo.15310859> (Schoenberg et al., 2025a). The full infrastructure to run scenario ensembles with FRIDA is hosted on GitHub <https://github.com/BenjaminBlanz/WorldTransFrida-Uncertainty>. EMB ensemble data is available <https://zenodo.org/records/15396799> (Schoenberg, 2025).
- Author contributions.* Conceptualisation: BC, MBG, WS. Data curation: MBG, BC, BB. Formal analysis: MBG, BB, BC, WS. Investigation: MBG, WS, BC, BB. Methodology: MBG, BC, WS, BB. Software: MBG, WS, BB. Supervision: WS, BC. Visualization: MBG, BB, WS. Writing - original draft: MBG. Writing - review and editing: BC, BB, WS, MBG.
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- ## References
- Acemoglu, D. and Cao, D.: Innovation by entrants and incumbents, *Journal of Economic Theory*, 157, 255–294, <https://doi.org/10.1016/j.jet.2015.01.001>, 2015.
- Ackerman, F., DeCanio, S. J., Howarth, R. B., and Sheeran, K.: Limitations of integrated assessment models of climate change, *Climatic Change*, 95, 297–315, <https://doi.org/10.1007/s10584-009-9570-x>, 2009.
- Adelman, I. and Yeldan, E.: The Minimal Conditions for a Financial Crisis: A Multiregional Intertemporal CGE Model of the Asian Crisis, *World Development*, 28, 1087–1100, [https://doi.org/10.1016/S0305-750X\(00\)00014-0](https://doi.org/10.1016/S0305-750X(00)00014-0), 2000.
- Adrian, T. and Shin, H. S.: Liquidity and Financial Cycles, *SSRN Journal*, <https://doi.org/10.2139/ssrn.1165583>, 2008.
- Aghion, P., Askenazy, P., Berman, N., Cetto, G., and Eymard, L.: Credit Constraints and the Cyclicity of R&D Investment: Evidence from France, *Journal of the European Economic Association*, 10, 1001–1024, <https://doi.org/10.1111/j.1542-4774.2012.01093.x>, 2012.
- Aghion, P., Akegiti, U., and Howitt, P.: What Do We Learn From Schumpeterian Growth Theory?, in: *Handbook of Economic Growth*, vol. 2, Elsevier, 515–563, <https://doi.org/10.1016/B978-0-444-53540-5.00001-X>, 2014.
- Ahmad, M.: Non-linear dynamics of innovation activities over the business cycles: Empirical evidence from OECD economies, *Technology in Society*, 67, 101721, <https://doi.org/10.1016/j.techsoc.2021.101721>, 2021.
- Alberti, M., Asbjornsen, H., Baker, L. A., Brozovic, N., Drinkwater, L. E., Drzyzga, S. A., Jantz, C. A., Fragoso, J., Holland, D. S., Kohler, T. (Tim) A., Liu, J. (Jack), McConnell, W. J., Maschner, H. D. G., Millington, J. D. A., Monticino, M., Podestá, G., Pontius, R. G., Redman, C. L., Reo, N. J., Sailor, D., and Urquhart, G.: Research on Coupled Human and Natural Systems (CHANS): Approach, Challenges, and Strategies, *Bulletin Ecologic Soc America*, 92, 218–228, <https://doi.org/10.1890/0012-9623-92.2.218>, 2011.
- Altonji, J. G. and Devereux, P. J.: The Extent and Consequences of Downward Nominal Wage Rigidity, <https://papers.ssrn.com/abstract=202741>, 1 July 1999.

- Alvarez-Cuadrado, F. and Poschke, M.: Structural Change Out of Agriculture: Labor Push versus Labor Pull, *American Economic Journal: Macroeconomics*, 3, 127–158, <https://doi.org/10.1257/mac.3.3.127>, 2011.
- 775 Andreyeva, T., Long, M. W., and Brownell, K. D.: The Impact of Food Prices on Consumption: A Systematic Review of Research on the Price Elasticity of Demand for Food, *Am J Public Health*, 100, 216–222, <https://doi.org/10.2105/AJPH.2008.151415>, 2010.
- Annicchiarico, B., Carattini, S., Fischer, C., and Heutel, G.: Business Cycles and Environmental Policy: A Primer, *Environmental and Energy Policy and the Economy*, 3, 221–253, <https://doi.org/10.1086/717222>, 2022.
- 780 Antonelli, C.: Endogenous innovation: the creative response, *Economics of Innovation and New Technology*, 26, 689–718, <https://doi.org/10.1080/10438599.2016.1257444>, 2017.
- Ardagna, S., Caselli, F., and Lane, T.: Fiscal Discipline and the Cost of Public Debt Service: Some Estimates for OECD Countries, *The B.E. Journal of Macroeconomics*, 7, <https://doi.org/10.2202/1935-1690.1417>, 2007.
- Auer, R. A., Levchenko, A. A., and Sauré, P.: International Inflation Spillovers through Input Linkages, *The Review of Economics and Statistics*, 101, 507–521, https://doi.org/10.1162/rest_a_00781, 2019.
- 785 Bartsch, F., Busies, I., Emambakhsh, T., Grill, M., Simoens, M., Spaggiari, M., and Tamburrini, F.: Designing a macroprudential capital buffer for climate-related risks, *European Central Bank*, LU, 2024.
- Battiston, S., Mandel, A., Monasterolo, I., Schütze, F., and Visentin, G.: A climate stress-test of the financial system, *Nature Clim Change*, 7, 283–288, <https://doi.org/10.1038/nclimate3255>, 2017.
- 790 Battiston, S., Monasterolo, I., Riahi, K., and van Ruijven, B. J.: Accounting for finance is key for climate mitigation pathways, *Science*, 372, 918–920, <https://doi.org/10.1126/science.abf3877>, 2021a.
- Battiston, S., Dafermos, Y., and Monasterolo, I.: Climate risks and financial stability, *Journal of Financial Stability*, 54, 100867, <https://doi.org/10.1016/j.jfs.2021.100867>, 2021b.
- Beck, R., Jakubik, P., and Piloiiu, A.: Key Determinants of Non-performing Loans: New Evidence from a Global Sample, *Open Econ Rev*, 26, 525–550, <https://doi.org/10.1007/s11079-015-9358-8>, 2015.
- 795 Beers, D. T. and Mavalwalla, J.: Database of Sovereign Defaults, 2017, SSRN Journal, <https://doi.org/10.2139/ssrn.3000226>, 2017.
- Bernanke, B. S. and Blinder, A. S.: Credit, Money, and Aggregate Demand, *The American Economic Review*, 78, 435–439, 1988.
- 800 Bhandari, J. S. and Weiss, L. A.: The Increasing Bankruptcy Filing Rate: An Historical Analysis, *Am. Bankr. L.J.*, 67, 1, 1993.
- Bohn, H.: The Behavior of U. S. Public Debt and Deficits, *The Quarterly Journal of Economics*, 113, 949–963, <https://doi.org/10.1162/003355398555793>, 1998.
- Brown, J. R., Fazzari, S. M., and Petersen, B. C.: Financing Innovation and Growth: Cash Flow, External Equity, and the 1990s R&D Boom, *The Journal of Finance*, 64, 151–185, <https://doi.org/10.1111/j.1540-6261.2008.01431.x>, 2009.

- 805 Brunetti, C., Dennis, B., Gates, D., Hancock, D., Ignell, D., Kiser, E. K., Kotta, G., Kovner, A., Rosen, R. J., and Tabor, N. K.: Climate Change and Financial Stability, FEDS Notes, 2021, <https://doi.org/10.17016/2380-7172.2893>, 2021.
- Burke, M. A. and Ozdagli, A.: Household Inflation Expectations and Consumer Spending: Evidence from Panel Data, *Review of Economics and Statistics*, 105, 948–961, https://doi.org/10.1162/rest_a_01118, 2023.
- 810 Carattini, S., Heutel, G., and Melkadze, G.: Climate policy, financial frictions, and transition risk, *Review of Economic Dynamics*, 51, 778–794, <https://doi.org/10.1016/j.red.2023.08.003>, 2023.
- Carney, M.: *Breaking the tragedy of the horizon—climate change and financial stability*, Lloyd’s of London, 2015.
- Carroll, C., Slacalek, J., Tokuoka, K., and White, M. N.: The distribution of wealth and the marginal propensity to consume: The distribution of wealth, *Quantitative Economics*, 8, 977–1020, <https://doi.org/10.3982/QE694>, 2017.
- 815 Cavana, R. Y.: *Feedback Economics: Economic Modeling with System Dynamics*, Springer International Publishing AG, Cham, 1 pp., 2021.
- Chinloy, P., Jiang, C., and John, K.: Investment, depreciation and obsolescence of R&D, *Journal of Financial Stability*, 49, 100757, <https://doi.org/https://doi.org/10.1016/j.jfs.2020.100757>, 2020.
- Chrysanthakopoulos, C., Konstantinou, P., and Tagkalakis, A.: Government spending cyclicality, economic stability and uncertainty, *Economic Systems*, 101314, <https://doi.org/10.1016/j.ecosys.2025.101314>, 2025.
- 820 Clarida, R., Galí, J., and Gertler, M.: Monetary policy rules in practice: Some international evidence, *European Economic Review*, 42, 1033–1067, [https://doi.org/10.1016/S0014-2921\(98\)00016-6](https://doi.org/10.1016/S0014-2921(98)00016-6), 1998.
- Coucke, K., Pennings, E., and Sleuwaegen, L.: Employee layoff under different modes of restructuring: exit, downsizing or relocation, *Industrial and Corporate Change*, 16, 161–182, <https://doi.org/10.1093/icc/dtm002>, 2007.
- 825 Cukierman, A. and Lippi, F.: Central bank independence, centralization of wage bargaining, inflation and unemployment:, *European Economic Review*, 43, 1395–1434, [https://doi.org/10.1016/S0014-2921\(98\)00128-7](https://doi.org/10.1016/S0014-2921(98)00128-7), 1999.
- Dasgupta, S., Van Maanen, N., Gosling, S. N., Piontek, F., Otto, C., and Schleussner, C.-F.: Effects of climate change on combined labour productivity and supply: an empirical, multi-model study, *The Lancet Planetary Health*, 5, e455–e465, [https://doi.org/10.1016/S2542-5196\(21\)00170-4](https://doi.org/10.1016/S2542-5196(21)00170-4), 2021.
- 830 Dell’Ariccia, G. and Marquez, R.: Lending Booms and Lending Standards, *The Journal of Finance*, 61, 2511–2546, <https://doi.org/10.1111/j.1540-6261.2006.01065.x>, 2006.
- Dembiermont, C., Scatigna, M., Szemere, R., and Tissot, B.: *A New Database on General Government Debt*, <https://papers.ssrn.com/abstract=2661592>, 13 September 2015.
- Deniz, P., Tekce, M., and Yilmaz, A.: Investigating the Determinants of Inflation: A Panel Data Analysis, *International Journal of Financial Research*, 7, p233, <https://doi.org/10.5430/ijfr.v7n2p233>, 2016.
- 835 Diamond, A.: Schumpeter’s Creative Destruction: A Review of the Evidence, *Journal of Private Enterprise*, 22, 120–146, 2006.
- Dietz, S., Bowen, A., Dixon, C., and Gradwell, P.: ‘Climate value at risk’ of global financial assets, *Nature Clim Change*, 6, 676–679, <https://doi.org/10.1038/nclimate2972>, 2016.

- 840 Doda, B.: Evidence on business cycles and emissions, *Journal of Macroeconomics*, 40, 214–227, <https://doi.org/10.1016/j.jmacro.2014.01.003>, 2014.
- Donges, J. F., Lucht, W., Cornell, S. E., Heitzig, J., Barfuss, W., Lade, S. J., and Schlüter, M.: Taxonomies for structuring models for World–Earth systems analysis of the Anthropocene: subsystems, their interactions and social–ecological feedback loops, *Earth Syst. Dynam.*, 12, 1115–1137, <https://doi.org/10.5194/esd-12-1115-2021>, 2021.
- 845 Dosi, G., Fagiolo, G., and Roventini, A.: Schumpeter meeting Keynes: A policy-friendly model of endogenous growth and business cycles, *Journal of Economic Dynamics and Control*, 34, 1748–1767, <https://doi.org/10.1016/j.jedc.2010.06.018>, 2010.
- Ehrlich, G. and Montes, J.: Wage Rigidity and Employment Outcomes: Evidence from Administrative Data, *American Economic Journal: Macroeconomics*, 16, 147–206, <https://doi.org/10.1257/mac.20200125>, 2024.
- 850 Elster, J.: *Explaining Social Behavior: More Nuts and Bolts for the Social Sciences*, 2nd ed., Cambridge University Press, Cambridge, <https://doi.org/10.1017/CBO9781107763111>, 2015.
- Fang, X., Hardy, B., and Lewis, K. K.: Who Holds Sovereign Debt and Why It Matters, *The Review of Financial Studies*, 38, 2326–2361, <https://doi.org/10.1093/rfs/hhaf031>, 2025.
- Farmer, J. D., Hepburn, C., Mealy, P., and Teytelboym, A.: A Third Wave in the Economics of Climate Change, *Environ Resource Econ*, 62, 329–357, <https://doi.org/10.1007/s10640-015-9965-2>, 2015.
- 855 Feldmann, H.: Technological unemployment in industrial countries, *J Evol Econ*, 23, 1099–1126, <https://doi.org/10.1007/s00191-013-0308-6>, 2013.
- Feng, F., Han, L., Jin, J., and Li, Y.: Climate Change Exposure and Bankruptcy Risk, *British J of Management*, 35, 1843–1866, <https://doi.org/10.1111/1467-8551.12792>, 2024.
- 860 Fishman, M. J., Parker, J. A., and Straub, L.: A Dynamic Theory of Lending Standards, *The Review of Financial Studies*, 37, 2355–2402, <https://doi.org/10.1093/rfs/hhae010>, 2024.
- Garcia-Jorcano, L. and Sanchis-Marco, L.: Measuring the impact of climate transition risk on the systemic risk: A multivariate quantile-located ES approach, *Research in International Business and Finance*, 80, 103127, <https://doi.org/10.1016/j.ribaf.2025.103127>, 2025.
- 865 García-Quevedo, J., Segarra-Blasco, A., and Teruel, M.: Financial constraints and the failure of innovation projects, *Technological Forecasting and Social Change*, 127, 127–140, <https://doi.org/10.1016/j.techfore.2017.05.029>, 2018.
- Ghosh, A. R., Kim, J. I., Mendoza, E. G., Ostry, J. D., and Qureshi, M. S.: Fiscal Fatigue, Fiscal Space and Debt Sustainability in Advanced Economies, *The Economic Journal*, 123, F4–F30, <https://doi.org/10.1111/econj.12010>, 2013.
- 870 Gillingham, K., Nordhaus, W., Anthoff, D., Blanford, G., Bosetti, V., Christensen, P., McJeon, H., and Reilly, J.: Modeling Uncertainty in Integrated Assessment of Climate Change: A Multimodel Comparison, *Journal of the Association of Environmental and Resource Economists*, 5, 791–826, <https://doi.org/10.1086/698910>, 2018.
- Giuzio, M., Krusec, D., Levels, A., Melo, A. S., Mikkonen, K., and Radulova, P.: Climate change and financial stability, European Central Bank, 2019.

- Gjeçi, A., Marinč, M., and Rant, V.: Non-performing loans and bank lending behaviour, *Risk Manag.* 25, 7, <https://doi.org/10.1057/s41283-022-00111-z>, 2023.
- 875 Godley, W. and Lavoie, M.: *Monetary Economics*, Palgrave Macmillan UK, London, <https://doi.org/10.1057/9780230626546>, 2007.
- Gompers, P., Kovner, A., Lerner, J., and Scharfstein, D.: Venture capital investment cycles: The impact of public markets, *Journal of financial economics*, 87, 1–23, 2008.
- Griffin, P., Lont, D., and Lubberink, M.: Extreme high surface temperature events and equity-related physical climate risk, *Weather and Climate Extremes*, 26, 100220, <https://doi.org/10.1016/j.wace.2019.100220>, 2019.
- 880 Grigsby, J., Hurst, E., and Yildirmaz, A.: Aggregate Nominal Wage Adjustments: New Evidence from Administrative Payroll Data, *American Economic Review*, 111, 428–471, <https://doi.org/10.1257/aer.20190318>, 2021.
- Gup, B. E.: *Too Big to Fail: Policies and Practices in Government Bailouts*, Bloomsbury Publishing USA, 368 pp., 2003.
- Hänsel, M. C., Drupp, M. A., Johansson, D. J. A., Nesje, F., Azar, C., Freeman, M. C., Groom, B., and Sterner, T.: Climate economics support for the UN climate targets, *Nat. Clim. Chang.*, 10, 781–789, <https://doi.org/10.1038/s41558-020-0833-x>, 2020.
- 885 Hasan, I. and Tucci, C. L.: The innovation–economic growth nexus: Global evidence, *Research Policy*, 39, 1264–1276, <https://doi.org/10.1016/j.respol.2010.07.005>, 2010.
- Huljak, I., Martin, R., Moccerro, D., and Pancaro, C.: Do non-performing loans matter for bank lending and the business cycle in euro area countries?, *Journal of Applied Economics*, 25, 1050–1080, <https://doi.org/10.1080/15140326.2022.2094668>, 2022.
- 890 ILO: Employment by sex and age -- ILO modelled estimates, Nov. 2024 (thousands) -- Annual, ILOSTAT [data set], https://doi.org/https://rshiny.ilo.org/dataexplorer15/?id=EMP_2EMP_SEX_AGE_NB_A&ref_area=X01&sex=SEX_T&classif1=AGE_YTHADULT_YGE15&timefrom=1991, 2024a.
- 895 ILO: Labour force participation rate by sex and age -- ILO modelled estimates, Nov. 2024 (%) -- Annual, ILOSTAT [data set], https://doi.org/https://rplumber.ilo.org/dataexplorer/?id=EAP_2WAP_SEX_AGE_RT_A&ref_area=X01&sex=SEX_T&classif1=AGE_YTHADULT_YGE15&timefrom=1990&timeto=2025, 2024b.
- ILO: SDG indicator 10.4.1 - Labour income share as a percent of GDP (%) -- Annual, ILOSTAT [data set], https://doi.org/https://rplumber.ilo.org/dataexplorer/?id=SDG_1041_NOC_RT_A&timefrom=2014&timeto=2024, 2024c.
- 900 ILO: Unemployment by sex and age -- ILO modelled estimates, Nov. 2024 (thousands) -- Annual, ILOSTAT [data set], https://doi.org/https://rshiny.ilo.org/dataexplorer16/?id=UNE_2UNE_SEX_AGE_NB_A&ref_area=X01&sex=SEX_T&classif1=AGE_YTHADULT_YGE15&timefrom=1991, 2024d.
- IPCC (Ed.): *Introduction and Framing*, in: *Climate Change 2022 - Mitigation of Climate Change*, Cambridge University Press, 151–214, <https://doi.org/10.1017/9781009157926.003>, 2023.
- 905 Jacobs, J., Ogawa, K., Sterken, E., and Tokutsu, I.: Public Debt, Economic Growth and the Real Interest Rate: A Panel VAR Approach to EU and OECD Countries, *Applied Economics*, 52, 1377–1394, <https://doi.org/10.1080/00036846.2019.1673301>, 2020.

- Jappelli, T. and Pistaferri, L.: The Consumption Response to Income Changes, *Annu. Rev. Econ.*, 2, 479–506, <https://doi.org/10.1146/annurev.economics.050708.142933>, 2010.
- Kahn, G. A.: International Differences in Wage Behavior: Real, Nominal, or Exaggerated?, *The American Economic Review*, 74, 155–159, 1984.
- Keppo, I., Butnar, I., Bauer, N., Caspani, M., Edelenbosch, O., Emmerling, J., Fragkos, P., Guivarch, C., Harmsen, M., Lefèvre, J., Le Gallic, T., Leimbach, M., McDowall, W., Mercure, J.-F., Schaeffer, R., Trutnevyte, E., and Wagner, F.: Exploring the possibility space: taking stock of the diverse capabilities and gaps in integrated assessment models, *Environ. Res. Lett.*, 16, 053006, <https://doi.org/10.1088/1748-9326/abe5d8>, 2021.
- Kiley, M. T.: Growth at Risk From Climate Change, *FEDS*, 2021, 1–19, <https://doi.org/10.17016/feds.2021.054>, 2021.
- Kortum, S. and Lerner, J.: Does Venture Capital Spur Innovation?, National Bureau of Economic Research, Cambridge, MA, <https://doi.org/10.3386/w6846>, 1998.
- Kraft, E. and Jankov, L.: Does speed kill? Lending booms and their consequences in Croatia, *Journal of Banking & Finance*, 29, 105–121, <https://doi.org/10.1016/j.jbankfin.2004.06.025>, 2005.
- Kramer, D. B., Hartter, J., Boag, A. E., Jain, M., Stevens, K., Nicholas, K. A., McConnell, W. J., and Liu, J.: Top 40 questions in coupled human and natural systems (CHANS) research, *E&S*, 22, art44, <https://doi.org/10.5751/ES-09429-220244>, 2017.
- Lamperti, F., Bosetti, V., Roventini, A., and Tavoni, M.: The public costs of climate-induced financial instability, *Nat. Clim. Chang.*, 9, 829–833, <https://doi.org/10.1038/s41558-019-0607-5>, 2019.
- Larkin, Y., Leary, M. T., and Michaely, R.: Do Investors Value Dividend-Smoothing Stocks Differently?, *Management Science*, 63, 4114–4136, <https://doi.org/10.1287/mnsc.2016.2551>, 2017.
- Liao, H.-H., Chen, T.-K., and Lu, C.-W.: Bank credit risk and structural credit models: Agency and information asymmetry perspectives, *Journal of Banking & Finance*, 33, 1520–1530, <https://doi.org/10.1016/j.jbankfin.2009.02.016>, 2009.
- Lim, Y. C. and Sek, S. K.: An Examination on the Determinants of Inflation, *JOEBM*, 3, 678–682, <https://doi.org/10.7763/JOEBM.2015.V3.265>, 2015.
- Liu, J., Dietz, T., Carpenter, S. R., Folke, C., Alberti, M., Redman, C. L., Schneider, S. H., Ostrom, E., Pell, A. N., Lubchenco, J., Taylor, W. W., Ouyang, Z., Deadman, P., Kratz, T., and Provencher, W.: Coupled Human and Natural Systems, *AMBIO: A Journal of the Human Environment*, 36, 639–649, [https://doi.org/10.1579/0044-7447\(2007\)36%5B639:CHANS%5D2.0.CO;2](https://doi.org/10.1579/0044-7447(2007)36%5B639:CHANS%5D2.0.CO;2), 2007.
- López-García, P., Montero, J. M., and Moral-Benito, E.: Business Cycles and Investment in Productivity-Enhancing Activities: Evidence from Spanish Firms, *Industry & Innovation*, 20, 611–636, <https://doi.org/10.1080/13662716.2013.849456>, 2013.
- Lown, C. and Morgan, D. P.: The Credit Cycle and the Business Cycle: New Findings Using the Loan Officer Opinion Survey, *Journal of Money, Credit and Banking*, 38, 1575–1597, 2006.
- Mandel, A., Battiston, S., and Monasterolo, I.: Mapping global financial risks under climate change, *Nat. Clim. Chang.*, 15, 329–334, <https://doi.org/10.1038/s41558-025-02244-x>, 2025.

- McLeay, M., Radia, A., and Thomas, R.: Money Creation in the Modern Economy, <https://papers.ssrn.com/abstract=2416234>, 14 March 2014.
- Meade, N. and Islam, T.: Modelling and forecasting the diffusion of innovation – A 25-year review, *International Journal of Forecasting*, 22, 519–545, <https://doi.org/10.1016/j.ijforecast.2006.01.005>, 2006.
- Meadows, D. H. and Wright, D.: Thinking in systems: a primer, Chelsea Green Pub, White River Junction, Vt, 218 pp., 2008.
- 950 Mendi, P.: Concentration of Innovation Investments Along the Business Cycle, *J Knowl Econ*, 15, 2856–2873, <https://doi.org/10.1007/s13132-023-01267-z>, 2024.
- Menicucci, E. and Paolucci, G.: The determinants of bank profitability: empirical evidence from European banking sector, *JFRA*, 14, 86–115, <https://doi.org/10.1108/JFRA-05-2015-0060>, 2016.
- 955 Mercure, J.-F., Knobloch, F., Pollitt, H., Paroussos, L., Scricciu, S. S., and Lewney, R.: Modelling innovation and the macroeconomics of low-carbon transitions: theory, perspectives and practical use, *Climate Policy*, 19, 1019–1037, <https://doi.org/10.1080/14693062.2019.1617665>, 2019.
- Minsky, H. P.: Stabilizing an unstable economy, McGraw-Hill Education, New York, 1 pp., 2008.
- Mortensen, D. T. and Nagypál, É.: More on unemployment and vacancy fluctuations, *Review of Economic Dynamics*, 10, 327–347, <https://doi.org/10.1016/j.red.2007.01.004>, 2007.
- 960 Nanda, R. and Rhodes-Kropf, M.: Investment cycles and startup innovation, *Journal of Financial Economics*, 110, 403–418, <https://doi.org/10.1016/j.jfineco.2013.07.001>, 2013.
- Narayan, P. K. and Narayan, S.: Government revenue and government expenditure nexus: evidence from developing countries, *Applied Economics*, 38, 285–291, <https://doi.org/10.1080/00036840500369209>, 2006.
- 965 NGFS: A Call for Action: Climate Change as a Source of Financial Risk, Network for Greening the Financial System, Paris, 2019.
- Nordhaus, W.: Climate Change: The Ultimate Challenge for Economics, *American Economic Review*, 109, 1991–2014, <https://doi.org/10.1257/aer.109.6.1991>, 2019.
- OECD: Benefits in unemployment, share of previous income, Data [dataset], 2025.
- 970 Ojea-Ferreiro, J., Reboredo, J. C., and Ugolini, A.: Systemic risk effects of climate transition on financial stability, *International Review of Financial Analysis*, 96, 103722, <https://doi.org/10.1016/j.irfa.2024.103722>, 2024.
- Parker, M. I.: Global Inflation: The Role of Food, Housing and Energy Prices, *SSRN Journal*, <https://doi.org/10.2139/ssrn.2923137>, 2017.
- Pindyck, R. S.: Climate Change Policy: What Do the Models Tell Us?, *Journal of Economic Literature*, 51, 860–872, <https://doi.org/10.1257/jel.51.3.860>, 2013.
- 975 Pindyck, R. S.: The Use and Misuse of Models for Climate Policy, *Review of Environmental Economics and Policy*, 11, 100–114, <https://doi.org/10.1093/reep/rew012>, 2017.

- Pollitt, H. and Mercure, J.-F.: The role of money and the financial sector in energy-economy models used for assessing climate and energy policy, *Climate Policy*, 18, 184–197, <https://doi.org/10.1080/14693062.2016.1277685>, 2018.
- 980 Poplawski-Ribeiro, M., Yoo, J., Haver, V., Kiendrebeogo, Y., Perrelli, R., Wei, Z., and Zhang, C.: *Global Debt Monitor 2023*, IMF, 2023.
- Powell, M. J. D.: *The BOBYQA Algorithm for Bound Constrained Optimization without Derivatives*, University of Cambridge, Cambridge, UK, 2009.
- 985 Qiao, Y., Dawson, A. R., Parry, T., and Flintsch, G. W.: Evaluating the effects of climate change on road maintenance intervention strategies and Life-Cycle Costs, *Transportation Research Part D: Transport and Environment*, 41, 492–503, <https://doi.org/10.1016/j.trd.2015.09.019>, 2015.
- Quatraro, F.: Co-evolutionary Patterns in Regional Knowledge Bases and Economic Structure: Evidence from European Regions, *Regional Studies*, 50, 513–539, <https://doi.org/10.1080/00343404.2014.927952>, 2016.
- Ram, R.: Additional Evidence on Causality between Government Revenue and Government Expenditure, *Southern Economic Journal*, 54, 763, <https://doi.org/10.2307/1059018>, 1988.
- 990 Ramey, V. A.: Macroeconomic shocks and their propagation, *Handbook of macroeconomics*, 2, 71–162, 2016.
- Ramme, L., Blanz, B., Wells, C., Wong, T. E., Schoenberg, W., Smith, C., and Li, C.: Feedback-based sea level rise impact modelling for integrated assessment models with FRISIAv1.0, <https://doi.org/10.5194/egusphere-2025-1875>, 27 June 2025.
- 995 Rennert, K., Errickson, F., Prest, B. C., Rennels, L., Newell, R. G., Pizer, W., Kingdon, C., Wingenroth, J., Cooke, R., Parthum, B., Smith, D., Cromar, K., Diaz, D., Moore, F. C., Müller, U. K., Plevin, R. J., Raftery, A. E., Ševčíková, H., Sheets, H., Stock, J. H., Tan, T., Watson, M., Wong, T. E., and Anthoff, D.: Comprehensive evidence implies a higher social cost of CO₂, *Nature*, 610, 687–692, <https://doi.org/10.1038/s41586-022-05224-9>, 2022.
- Rodano, G., Serrano-Velarde, N., and Tarantino, E.: Lending Standards over the Credit Cycle, *The Review of Financial Studies*, 31, 2943–2982, <https://doi.org/10.1093/rfs/hhy023>, 2018.
- 1000 Saltelli, A. (Ed.): *Global sensitivity analysis: the primer*, John Wiley, Chichester, England Hoboken, NJ, 1 pp., <https://doi.org/10.1002/9780470725184>, 2008.
- Sanders, M., Serebriakova, A., Fragkos, P., Polzin, F., Egli, F., and Steffen, B.: Representation of financial markets in macro-economic transition models—a review and suggestions for extensions, *Environ. Res. Lett.*, 17, 083001, <https://doi.org/10.1088/1748-9326/ac7f48>, 2022.
- 1005 Schoenberg, W.: FRIDA v2.1 Endogenous Model Behavior (EMB) 100000 member ensemble (1.0.0), <https://doi.org/10.5281/ZENODO.15396798>, 2025.
- Schoenberg, W., Blanz, B., Ramme, L., Wells, C., Grimeland, M., Callegari, B., Breier, J., Rajah, J., Nicolaidis Lindqvist, A., Mashhadi, S., Muralidhar, A., and Eriksson, A.: FRIDA: Feedback-based knowledge Repository for Integrated Assessments (v2.1), <https://doi.org/10.5281/ZENODO.15310859>, 2025a.
- 1010 Schoenberg, W., Blanz, B., Rajah, J. K., Callegari, B., Wells, C., Breier, J., Grimeland, M. B., Lindqvist, A. N., Ramme, L., Smith, C., Li, C., Mashhadi, S., Muralidhar, A., and Mauritzen, C.: Introducing FRIDA v2.1: A feedback-based, fully coupled, global integrated assessment model of climate and humans, <https://doi.org/10.5194/egusphere-2025-2599>, 26 June 2025b.

- Schumpeter, J. A.: The theory of economic development; an inquiry into profits, capital, credit, interest, and the business cycle, Harvard University Press, Cambridge, Mass., 1934.
- 1015 Schumpeter, J. A.: Business Cycles: A Theoretical, Historical and Statistical Analysis of the Capitalist Process, McGraw-Hill Book Co., New York, 1939.
- Schumpeter, J. A.: Socialism, Capitalism and Democracy, Harper and Brothers, New York, 1942.
- Solow, R. M. and Taylor, J. B. (Eds.): Inflation, unemployment, and monetary policy, The MIT Press, Cambridge, Mass, 120 pp., 1998.
- 1020 Stern, N.: Economics: Current climate models are grossly misleading, *Nature*, 530, 407–409, <https://doi.org/10.1038/530407a>, 2016.
- Stern, N.: Towards a carbon neutral economy: How government should respond to market failures and market absence, *Journal of Government and Economics*, 6, 100036, <https://doi.org/10.1016/j.jge.2022.100036>, 2022.
- 1025 Stern, N., Stiglitz Charlotte Taylor, J., and Taylor, C.: The economics of immense risk, urgent action and radical change: towards new approaches to the economics of climate change, *Journal of Economic Methodology*, 29, 181–216, <https://doi.org/10.1080/1350178X.2022.2040740>, 2022.
- Science in the UNFCCC negotiations: <https://unfccc.int/topics/science/the-big-picture/science-in-the-unfccc-negotiations>, last access: 12 September 2025.
- 1030 van der Veer, K. J. M. and Hoeberichts, M. M.: The level effect of bank lending standards on business lending, *Journal of Banking & Finance*, 66, 79–88, <https://doi.org/10.1016/j.jbankfin.2016.01.003>, 2016.
- Wells, C., Blanz, B., Ramme, L., Breier, J., Callegari, B., Muralidhar, A., Rajah, J. K., Lindqvist, A. N., Eriksson, A. E., Schoenberg, W. A., Köberle, A. C., Wang-Erlandsson, L., Mauritzen, C., and Smith, C.: The Representation of Climate Impacts in the FRIDAv2.1 Integrated Assessment Model, <https://doi.org/10.5194/egusphere-2025-2756>, 14 July 2025.
- 1035 WID: Pre-tax national income | bottom 40% | share | adults | equal split | world, WID [data set], https://doi.org/https://wid.world/data/#countriestimeseries/sptinc_p0p40_z/WO/1820/2023/eu/k/p/yearly/s, 2024.
- Wong, P. K., Ho, Y. P., and Autio, E.: Entrepreneurship, Innovation and Economic Growth: Evidence from GEM data, *Small Bus Econ*, 24, 335–350, <https://doi.org/10.1007/s11187-005-2000-1>, 2005.
- World Bank: GDP, PPP (constant 2021 international \$), World Bank [data set], <https://doi.org/https://data.worldbank.org/indicator/NY.GDP.MKTP.PP.KD>, 2023a.
- 1040 World Bank: GDP, PPP (current international \$), World Bank [data set], <https://doi.org/https://data.worldbank.org/indicator/NY.GDP.MKTP.PP.CD>, 2023b.
- World Bank: Gross capital formation (% of GDP), World Bank [data set], <https://doi.org/https://data.worldbank.org/indicator/NE.GDI.TOTL.ZS>, 2023c.
- 1045 World Bank: Households and NPISHs final consumption expenditure (% of GDP), World Bank [data set], <https://doi.org/https://data.worldbank.org/indicator/NE.CON.PRVT.ZS>, 2023d.