1	Growth in Agricultural Water Demand Aggravates Water Supply-
2	Demand Risk in Arid Northwest China: More a result of
3	Anthropogenic Activities than Climate Change
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11	Highlights:
12	Irrigation water demand surge critically amplifies water supply-demand risk in arid
13	regions;
14	Water resources in arid regions are more susceptible to anthropogenic impacts;
15	Regional water supply-demand risk continues to rise through the mid-21st century.
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17	Abstract
18	The dynamic evolution mechanism of regional water supply-demand risks under
19	the combined effects of climate change and human activities remains unclear,
20	particularly against the backdrop of agricultural expansion in arid regions. This study
21	focuses on the Tailan River Basin (TRB), a typical arid watershed in China and a vital
22	base for high-quality fruit and grain production. By integrating the PLUS (Land Use
23	Simulation) and InVEST (Integrated Valuation of Ecosystem Services and Tradeoffs)
24	models, we constructed a water supply-demand risk assessment framework
25	encompassing 24 climate-land change scenarios to quantify their impacts on regional

26	water resource patterns and risks. Results reveal that that climate change profoundly	设置了格式: 字体颜色: 红色
27	influences water supply, while land use significantly affects water demand. Under the	设置了格式: 字体颜色: 红色
28	Balanced Economic and Ecological Development Scenario (BES), 531.2 km² of	
29	additional cultivated land could be developed by 2050. However, this cultivated land	设置了格式:字体颜色: 红色 设置了格式:字体颜色: 红色
30	expansion leads to a sharp increase in irrigation water demand, with the minimum	(以且 J 竹八. 于 P 例 巴. 红 巴
31	demand reaching 4.87×10 ⁸ m³, while the maximum regional water supply is only	
32	0.16×10 ⁸ m³, resulting in a significant supply-demand gap (>4.71×10 ⁸ m³). The risk	
33	assessment framework indicates that by 2050, the entire TRB will face a water supply-	
34	demand crisis, with at least 46% of the area experiencing severe (Level III) or higher	
35	risks. The study demonstrates that continuous cultivated land expansion driven by	设置了格式:字体颜色: 红色 设置了格式: 字体颜色: 红色
36	agricultural activities—which drastically increases irrigation water demand—is the root	KEJIIA J POKO. AL
37	cause of intensifying water supply-demand conflicts and high risks in the TRB. By 2050,	
38	the proportion of irrigation water to total water use will exceed 70%, regardless of	
39	scenario. These findings underscore the necessity of deeply integrating	设置了格式: 字体颜色: 红色
40	multidisciplinary approaches within a water risk framework to elucidate land-eco-	
41	hydrological feedback mechanisms and better address water security challenges under	
42	climate change. The results provide a scientific basis for optimizing regional water-land	
43	resource allocation and promoting agro-ecological sustainable	
44	development. Maintaining regional water supply-demand balance is crucial for	
45	achieving sustainable development. Although the impacts of climate change and	
46	anthropogenic activities on water resources are widely recognized, the dynamic	
47	response mechanisms of water supply demand risk (WSDR) under their combined	

forcing remain unclear. The Tailan River Basin (TRB), though situated in a typical arid climate zone, serves as China's vital fruit and high-quality grain production base due to abundant solar-thermal resources. However, systematic research on WSDR in this region is deficient, impeding guidance for healthy and stable development of largescale farming. To address this, we developed a WSDR analytical framework based on PLUS InVEST models, encompassing 24 climate land change scenarios. This framework quantifies impacts of climate change and anthropogenic activities on TRB's water supply demand patterns and associated risks. Results show that under the Balanced Economy-Ecology Strategy (BES), effective land consolidation could add 531.2 km² of cultivated land by 2050. However, significant cultivated land expansion drives minimum water demand to surge to 4.87×108 m³, while maximum regional water supply reaches only 0.16×108 m³, breaking the supply-demand balance. By 2050, the entire TRB will face WSDR crises, with at least 46% of the region enduring endangered (Level III) risk. The root cause is persistent anthropogenic activities particularly landuse change-triggering continuous cultivated land expansion, increasing irrigation water demand and intensifying conflicts between water demand and supply capacity. These findings underscore the need to deeply integrate multidisciplinary approaches within WSDR frameworks, in depth analysis of land ecology hydrology feedback mechanisms, to better address water security challenges under climate change. This study can provide an important scientific basis for the optimal allocation of regional soil and water resources and the sustainable development of agroecology.

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- **Keywords**: Climate change; Anthropogenic activities; Land use; Water supply-
- 71 demand risk (WSDR); Sustainable water governance

Introduction

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Arid zones, covering 41% of the Earth's land area, are critical components of global terrestrial ecosystems. They not only support 38% of the world's population but also host approximately one-third of the planet's biodiversity hotspots (Berdugo et al., 2017; Li et al., 2021). However, these regions—predominantly located in developing countries (Huang et al., 2016; Chen et al., 2024)—face extreme water scarcity and exhibit high ecological fragility (Li et al., 2021), making them particularly sensitive to human activities (especially agricultural practices) and climate change. Northwestern China's arid regions serve as a typical example of the interplay between ecological vulnerability and agricultural pressure. Since 1980, cultivated land in this area has expanded significantly by 25.87% (Zhu et al., 2021), profoundly altering water and land resource allocation and ecological balance (Liu et al., 2025b). Although climate change has led to increased runoff (Li et al., 2025b) and rainfall (Yao et al., 2022) in the region, providing more available water resources (Chen et al., 2023a), agricultural activities dominated by continuous cultivated land expansion have sharply intensified regional water stress. Irrigation water use has now become the major consumer of water resources. Simultaneously, cultivated land expansion has elevated evapotranspiration levels (Zhu et al., 2025), and inefficient irrigation practices (e.g., the irrigation water use efficiency in Xinjiang is only 0.585) have further exacerbated groundwater overextraction (Yan et al., 2025) and soil salinization (Perez et al., 2024). These factors have intensified the contradiction between water supply and demand, continuously constraining sustainable water use options, and amplifying ecological vulnerability (Huggins et al., 2022) and food security risks (Jones et al., 2024). Water deficits (supplydemand gaps) are squeezing the viability of sustainable water use and water security, exacerbating ecological vulnerability (Huggins et al., 2022) and food security risks (Jones et al., 2024), particularly in arid zones. Over recent decades, intensified climate change (Carr et al., 2024), increased frequency of extreme weather events (Van et al., 2023), persistent population growth (Dolan et al., 2021), and lagging water management (Zhang et al., 2025c) have collectively accelerated global water depletion, leaving approximately 4 billion people under severe water scarcity (Mekonnen et al., 2016). Mismatches and mismatches between natural endowments of water resources and human needs in terms of spatial distribution and temporal variations further exacerbate regional water scarcity problems and failure to meet ecosystem and societal needs (Caretta et al., 2024). Projections indicate that over the next three decades, population growth, rising food demand, and accelerated urbanization will drive sustained increases in multisectoral water withdrawals—particularly irrigation water (Flörke et al., 2018; He et al., 2021). The resulting surge in demand will lead to a chronic water deficit, posing a serious challenge to the achievement of the United Nations Sustainable Development Goals (SDGs), in particular SDG 6 (sustainable water management), SDG 7 (sustainable modern energy) and SDG 17 (sustainable use of terrestrial ecosystems) (Ma et al., 2025).

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The imbalance between water supply and demand is influenced by both climate change and human activities (particularly agricultural practices). Climate change profoundly affects key processes in the hydrological cycle, including alterations in precipitation and evapotranspiration (Konapala et al., 2020). The AR6 Synthesis Report highlights that for every 0.5°C increase in global temperature, extreme heatwaves, heavy rainfall, and regional droughts become more frequent and severe (Mukherji et al., 2023), elevating risks of extreme floods (surplus) and droughts (deficit). Research indicates that changes in critical climate variables (precipitation, temperature, evapotranspiration) significantly disrupt runoff patterns and alter the availability of surface water resources (Lipczynska et al., 2018). Simultaneously, agricultural activities (e.g., irrigation) directly impact the water cycle by modifying hydrological processes such as evaporation, soil moisture, and water storage, while also affecting water and energy balances through artificially enhanced evaporation (Yan et al., 2025). Furthermore, agricultural activities directly shape water supply and demand by altering water use patterns and intensity, thereby creating bidirectional feedback loops with the water cycle and ecosystems (Chen et al., 2023). Under the influence of climate change and agricultural activities, the mismatch between the natural endowment of water resources (in terms of spatiotemporal distribution) and human demands further exacerbates regional water scarcity, making it increasingly challenging to meet both 设置了格式: 字体颜色: 红色

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This mismatch and dislocation are jointly driven by climate change and human activities (particularly agricultural practices). Studies have demonstrated that the increased runoff observed during the 20th century resulted from the combined effects of climate change and land cover changes (Piao et al., 2007). Land use changes can influence precipitation patterns through modifications in surface energy balance, hydrological cycles, and large-scale atmospheric circulation (Zhang et al., 2025a), while climate change exacerbates the impacts of land alterations by reshaping the hydrological cycle, thereby aggravating meteorological extremes (e.g., floods and droughts). Furthermore, the relative influences of climate change and agricultural activities vary significantly across different environmental issues. Climate change dominates changes in runoff (Zeng et al., 2024), ecosystem services (Jia et al., 2024), and vegetation dynamics (Hu et al., 2025). In contrast, land use changes exert greater impacts than climate change on terrestrial productivity (He et al., 2025), carbon use efficiency (Chen et al., 2024b), and soil variables (Ding et al., 2024). However, the relative contributions of climate change and land use to water supply-demand balance, as well as how their interactions shape the spatial patterns and temporal evolution of supply-demand risks, remain poorly understood. Existing studies on water supply and demand have predominantly focused on unilateral impacts of either land use (Deng et al., 2024; Wen et al., 2025) or climate change (Gharib et al., 2023; Li et al., 2024). Simultaneously considering the effects of both climate and land use changes on water supply-demand balance is crucial and necessary (Liu et al., 2022; Guo et al., 2023). Therefore, investigating the response mechanisms of water supply-demand balance and risks under the combined effects of climate change and agricultural activities represents a critical scientific question that urgently needs to be addressed. Water resource surpluses and deficits are profoundly influenced by both climate change and human activities. Climate change deeply affects multiple key processes within the hydrological cycle, including changes in precipitation and evapotranspiration (Konapala et al., 2020). The AR6 Synthesis Report states that with every 0.5°C increase in global temperature, extreme heatwaves, heavy rainfall, and regional droughts become more frequent and

intense (Mukherji et al., 2023), elevating the risks of extreme flooding (surplus) and drought (deficit). Research shows that changes in key climatic variables (precipitation, temperature, evapotranspiration) significantly perturb runoff, altering the availability of surface water resources. These changes also affect the biodegradation of soil organic matter and pollutants in rivers (Lipezynska et al., 2018), limiting the available water volume in river systems. The impacts of human activities on the global terrestrial hydrological cycle are also multifaceted. Human activities have altered 41% of the land surface cover, and this alteration can trigger changes in the ratio of evaporation to runoff (Bosmans et al., 2017). Furthermore, land use/cover change can significantly alter canopy interception, soil infiltration, and evapotranspiration, consequently exerting significant impacts on runoff volume and peak flow (Guo et al., 2020). It also severely affects critical watershed hydrological processes and soil moisture parameters (such as surface roughness, infiltration rate, soil structure, and hydraulic conductivity) (Patra et al., 2023). These alterations further affect processes such as precipitation infiltration, groundwater recharge, and soil water storage, ultimately impacting regional water resource availability. Beyond altering the natural water cycle, human activities also directly impact water supply and demand by modifying water use structure and intensity, thereby inducing positive or negative feedbacks on the water cycle and ecosystems (Boakes et al., 2024).

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Model prediction serves as a powerful tool for analyzing land use changes, water resource evolution, and water supply-demand dynamics. The Patch-generating Land Use Simulation (PLUS) model, which integrates spatial, empirical, and statistical approaches, enables accurate analysis of drivers behind land use changes and patch evolution (Liang et al., 2021). Studies demonstrate that PLUS outperforms many other models in simulation precision, more realistically capturing the spatial characteristics of land use changes (Gao et al., 2022). The InVEST model excels in allocating water resources and evaluating water conservation functions at the watershed scale, offering advantages such as minimal data requirements and strong spatial representation capabilities. Its water yield module has been widely applied and validated for water

supply assessments across diverse global basins (Chen et al., 2024a; Ma et al., 2024). The coupled PLUS-InVEST framework has been extensively utilized in fields such as carbon storage simulation, habitat quality assessment, and optimization of ecosystem service spatial patterns (Zhang et al., 2024; Huang et al., 2024; Wang et al., 2024b). However, its application to deeply explore regional water supply-demand dynamics under climate change and irrigation agriculture remains limited. The impacts of climate change and human activities on water resources are not independent. Research indicates that the increase in runoff during the 20th century resulted from the combined effects of climate change and human activities, with the impact of human activities (specifically land cover change) even exceeding that of climate change in certain regions (Piao et al., 2007). Land use change can influence precipitation through alterations in surface energy balance, the water cycle, and large-scale atmospheric circulation (Zhang et al., 2025a). Climate change catalyzes and amplifies the effects of land alteration by modifying the hydrological cycle, thereby altering meteorological elements and triggering more severe extreme disasters (floods and droughts). Furthermore, the relative influence of climate change versus human activities varies significantly across different environmental element. Studies show that climate change dominates the changes in runoff (Zeng et al., 2024), ecosystem services (Jia et al., 2024), and vegetation dynamics (Hu et al., 2025). Whereas, land use exerts a greater influence than climate change on terrestrial productivity (He et al., 2025), carbon use efficiency (Chen et al., 2024b), and soil variables (Ding et al., 2024). However, there is still a lack of systematic understanding regarding the relative contributions of climate change and human activities to water resource supply-demand balance, and how their interactions shape the spatial patterns and temporal evolution trends of supply demand risks. Therefore, in-depth investigation into the response mechanisms of water resource supply demand balance and risks under the combined effects of climate change and human activities constitutes a critical scientific problem urgently requiring resolution.

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Based on this, our study focuses on a typical watershed in the arid region of Northwest China, aiming to investigate water supply-demand balance under the

influence of climate change and human activities, and to identify the primary factors driving water supply-demand risks. The specific objectives of this research are: i) To determine land change trends under six development scenarios (NIS, FSS, EDS, WPS, EPS, and BES) using the PLUS model, and to identify high-contribution factors driving land changes; ii) To clarify the dynamics of water supply and demand under 24 landclimate combination scenarios (incorporating four climate change scenarios and six land use change scenarios), and to analyze the key drivers behind these changes; iii) To quantify water supply-demand risks under these land-climate combination patterns and identify the main factors influencing these risks. By coupling multi-scenario analyses of climate and land use changes, this study systematically evaluates their impacts on water supply-demand patterns and associated risks in a typical arid basin, providing actionable recommendations for optimizing water-land resource allocation and promoting agro-ecological sustainable development in the region. Model prediction serves as a powerful tool for analyzing land use change, water resource evolution processes, and changes in water supply and demand, among others. The Patchgenerating Land Use Simulation (PLUS) model, which integrates spatial, empirical, and statistical models, can be used to accurately analyze land use change and patch growth (Liang et al., 2021). Research demonstrates that PLUS outperforms many other models in terms of simulation accuracy and captures the spatial characteristics of land use change more realistically (Gao et al., 2022). The InVEST model excels in allocating water resources and evaluating watershed water conservation functions, among other applications, while also boasting advantages such as low data requirements and strong spatial representation capabilities. Its water yield module has been widely applied and validated for water resource supply assessment across diverse watersheds globally (Chen et al., 2024a; Ma et al., 2024). Coupling these two models (PLUS and InVEST) has been extensively applied in fields such as carbon storage simulation, habitat quality assessment, and spatial optimization of ecosystem services (Zhang et al., 2024; Huang et al., 2024; Wang et al., 2024b). However, the application of the coupled PLUS-InVEST model to deeply investigate the response mechanisms of regional water supply and demand to the combined effects of climate change and human activities remains

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Water resources in arid regions are extremely scarce and their ecosystems are highly fragile (Li et al., 2021), making them highly sensitive to both climate change and human activities. Since 1980, cultivated land in China's arid regions has expanded significantly by 25.87% (Zhu et al., 2021), profoundly impacting the allocation of water and land resources and the ecological balance in these areas (Liu et al., 2025b). Under the influence of climate change, both runoff (Li et al., 2025b) and precipitation (Yao et al., 2022) have shown increasing trends, providing more available water resources for the region (Chen et al., 2023a). However, under the influence of human activities, the continuous expansion of cultivated land has led to a surge in regional water use pressure, with irrigation water now constituting the primary consumptive portion of water resources. Simultaneously, cultivated land expansion intensifies regional evapotranspiration (Zhu et al., 2025), and inefficient irrigation (with an irrigation water use efficiency coefficient of only 0.585 in Xinjiang) exacerbates groundwater overdraft (Yan et al., 2025) and soil salinization (Perez et al., 2024), continually intensifying the conflict between water supply and demand in arid regions. Although current research has focused on the individual impacts of climate change (Lu et al., 2024a; Hamed et al., 2024) and human activities (Zhu et al., 2023; Valjarević et al., 2022) on arid regions, there remains a lack of systematic investigation into the dynamic mechanisms of supply-demand risks driven by the coupling of these two factors. Therefore, within a comprehensive framework considering both climate change and human activities, it is necessary to quantify water supply demand risks in arid regions and identify the key controlling factors, in order to support sustainable water management strategies for these areas.

Based on this, we studied a typical watershed in the arid region of Northwest China, aiming to investigate water resource supply demand balance under the influence of climate change and human activities, and to analyze the main influencing factors of water supply demand risks. The specific objectives of this study are: i) To determine land change trends using the PLUS model under six development scenarios (NIS, FSS, EDS, WPS, EPS, and BES), and to identify the high contribution factors driving land

change. ii) To clarify the change processes of water supply and demand quantities under 24 land climate combination patterns (resulting from four climate change scenarios and six land use change scenarios), and to analyze the key reasons driving these water supply demand changes; iii) To quantify water supply demand risks under the land-climate combination patterns, and to identify the main factors influencing these water supply demand risks; iv) To propose practical management strategies and recommendations for water and land resource planning and allocation, and for agroecological sustainable development in typical arid region watersheds.

2 Datasets and methods

2.1 Study Area

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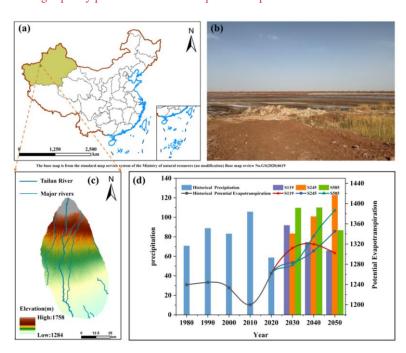
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The Tailan River originates from the southern foothills of Tomur Peak in the Tianshan Mountains and is primarily recharged by alpine snow and ice melt, with a multi-year average runoff of 7.766×108 m³. The Tailan River Basin (TRB) (Fig. 1) is a typical inland river basin in the arid region of northwestern China, covering a total area of 4,218 km². The basin features diverse landforms including gravel Gobi, alluvial plains, and fine soil plains, and is characterized by a continental arid climate of the northern temperate zone with intense solar radiation, high evaporation rates, an average annual precipitation of only 177.7 mm, and evaporation reaching 2,912 mm. The mean annual temperature is 8.6°C, with an average wind speed of 1.25 m/s (Fig. S1). Located in southern Xinjiang, TRB's climatic and hydrological characteristics are highly representative of arid regions both in China and globally. The process of water resources formation in its high mountain areas and consumption in the oasis-desert zones reflects the universal water cycle and utilization patterns of inland river basins in arid regions. TRB has a relatively concentrated population and developed oasis agriculture, forming a diversified agricultural production structure dominated by cotton and food crops, alongside equally important forestry and fruit industries, making it a typical representative of oasis economic systems in arid regions. As an important regional producer of grain, cotton, oil, and fruits, TRB yields high-quality rice and cotton, as well as abundant walnuts, apples, red dates, and fragrant pears. Its water and land resource utilization patterns and oasis-desert ecosystem structure provide valuable

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references for other arid river basins. Therefore, although TRB is a single basin, its physiographic conditions, climatic and hydrological characteristics, ecological structure, and human activity patterns all reflect the universal attributes of inland river basins in arid regions, possessing both typicality and representativeness for regional pattern studies. The Tailan River originates on the southern slope of Mount Tomur in the Tianshan Mountains. It is a typical inland river in the arid region of Northwest China (Fig. 1), primarily fed by glacial and snow meltwater, with a multi-year average runoff of 7.766×108 m³. The Tailan River Basin (TRB) is composed of gravel, alluvial, and fine soil plains. It experiences a continental north temperate arid climate, characterized by abundant sunshine and high evaporation. The multi-year average precipitation is 177.7 mm, evaporation is 2912 mm, air temperature is 8.6 °C, and wind speed is 1.25 m/s. The TRB is an important base for grain, cotton, oil crops, and fruits, having developed a diversified pattern centered on grain and cotton. It serves both as a trade hub for melons and fruits such as walnuts, apples, jujubes (red dates), and fragrant pears, and as a high quality production area for important crops like cotton and rice.



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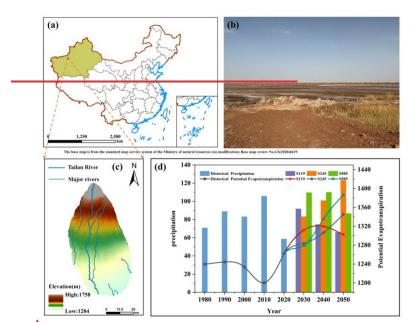


Fig. 1. Overview of the Tailan River Basin (TRB): (a) Schematic map showing the location of TRB in China; (b) Actual landscape of the Tailan River; (c) Digital Elevation Model (DEM) of

TRB; (d) Precipitation and potential evapotranspiration for historical and future periods in TRB.

2.2 Datasets

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This study collected two sets of datasets to simulate land use and water supplydemand in the TRB (Tab. 1). The first set of data was used to simulate land use change (Tab. 1), involving a total of 19 factors influencing land use to establish a driving factor library. These include 10 socio-economic factors, 3 climate factors, 3 topographic factors, 2 soil factors, and 1 vegetation factor. The second set of data was used to simulate water supply and demand quantities, with a total of 12 factors employed for the simulation (Tab. 2). Additionally, land use and future climate were used as the base data, and land use data were obtained from RESDC (https://www.resdc.cn/), constructed using interactive visual interpretation methods based on Landsat MSS,

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TM/ETM and Landsat 8 images (Zhuang et al., 1999), which include cultivated land, forest land, grassland, water bodies, built-up land and unutilized land, with an overall accuracy of more than 95% (Liu et al., 2014). Future meteorological data were obtained from TPDC (https://www.tpdc.ac.cn/), and Coupled Model Intercomparison Project (CMIP6) was selected as the data source. Considering the size of the study area, modeling efficiency, and information richness, bilinear interpolation was employed to harmonize the spatial resolution of all datasets to 30 meters within the Krasovsky_1940_Albers coordinate system.

Tab 1. Data Factors of the Land-Climate Model in the Tailan River Basin 设置了格式: 字体颜色: 红色

<u>Model</u>	<u>Category</u>	<u>Data</u>	<u>Year</u>	Spatial Resolution	Source
		Average annual precipitation	2000–2020		https://www.resdc.cn/
	Climatic	Average annual temperature	<u>2000–2020</u>	<u>1000 m</u>	nttps://www.resuc.cn/
		<u>Drought Index</u>	<u>2022</u>		https://www.plantplus.cn/
		<u>Digital Elevation Model</u>			
	<u>Terrain</u>	Slope	=	30 m	https://www.gscloud.cn/
		Slope direction		<u></u>	
	Soil	Soil type	2009		https://www.fao.org/
		Soil erosion	<u>2019</u>	<u>1000 m</u>	
PLUS	<u>Plant</u>	Normalized Difference		<u>30 m</u>	https://www.resdc.cn/
		<u>Vegetation Index</u>	2010, 2015,		
		<u>Population</u>	<u>2020</u>	<u>100 m</u>	https://hub.worldpop.org
		Gross Domestic Product		<u>1000 m</u>	https://www.resdc.cn/
		Nighttime lights		<u>500 m</u>	https://eogdata.mines.edu
	Socio-	Distance to railway			
	economic	Distance to highway	2020	20	1
		Distance to river system	<u>2020</u>	<u>30 m</u>	https://www.ngcc.cn/
		Distance to primary road Distance to secondary road			

		Distance to township Road			
		Distance to residential areas			
	CMIDO	Monthly precipitation			
	CMIP6	Monthly temperature	<u>2021–2100</u>	<u>30 m</u>	144 // 4 1
	(MRI-	Monthly potential			https://www.tpdc.ac.cn/
	ESM2.0)	evapotranspiration			
		Plant available water content			https://www.fao.org/
	<u>Soil</u>	Root restriction layer depth	2009		
InVEST	<u>5011</u>	Per-capita household water	<u>2009</u>	Ξ	<u>Yan (2020)</u>
IIVEST		consumption			
		Water consumption per 10,000			Xinjiang Uygur Autonomous
		¥ GDP			Region Water Resources
	Socioecon	Per-hectare farmland irrigation	<u>2000-2020</u>	Ξ.	<u>Bulletin</u>
	omic	consumption	2000 2020	_	<u>swieni.</u>
		GDP of Tailan River Basin			WenSu County and the Aksu
		POP of Tailan River Basin			City Statistical Yearbooks.
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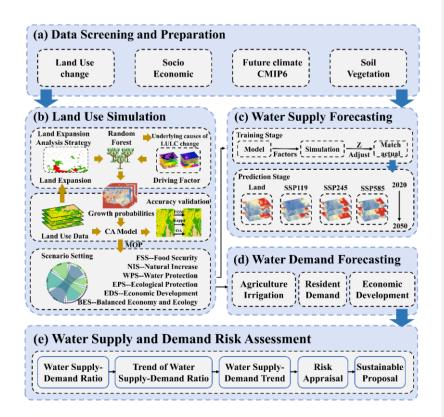
	Average annual precipitation	2000-2020		https://www.resde.en/
limatic	Average annual temperature	2000 2020	1000 m	
	Drought Index	2022		https://www.plantplus.cn
	Digital Elevation Model			
errain	Slope	-	30 m	https://www.gseloud.en
	Slope direction			
bil	Soil type	2009		https://www.fao.org/
011	Soil erosion	2019	1000 m	nups.//www.tuo.org/
lant	Normalized Difference		30 m	https://www.resde.en/
luint	Vegetation Index			nttps://www.restec.en/
	Population	2010, 2015, 2020	100 m	https://hub.worldpop.org
	Gross Domestic Product	2010, 2013, 2020	1000 m	https://www.resde.en/
	Nighttime lights		500 m	https://eogdata.mines.edu/
	regitaine ngna			ducts/vnl/
ocio-	Distance to railway			
onomic	Distance to highway			
ononne	Distance to river system			
	Distance to primary road	2020	30 m	https://www.ngcc.cn/
	Distance to secondary road			
	Distance to township Road			
	Distance to residential areas			

Data Year **Category Source** Monthly precipitation CMIP6 2021 2100 https://www.tpdc.ac.cn/ Monthly temperature (MRI ESM2.0) Plant available water content https://www.fao.org/ 2009 Soil Root restriction layer depth Yan (2020)

	Per capita household water		
	consumption		V:-:: II A4
	Water consumption per		Xinjiang Uygur Autonomous
	Water consumption per		Region Water Resources
	10,000 ¥ GDP		region water resources
Socioeconomic	10,000 1 321	2000-2020	Bulletin
	Per hectare farmland irrigation		
	consumption		
	GDP of Tailan River Basin		WenSu County and the Aksu
	ODF OF Tanan Kiver Basin		wensu County and the Aksu
	POP of Tailan River Basin		City Statistical Yearbooks.
			,

2.3 Methods

The research approach of this study is to first predict land use change in the TRB under six scenarios for the period 2020-2050 and screen for the high-contribution drivers of land change in the TRB. subsequently predict the change processes of water supply and demand quantities in the TRB under 24 land-climate combination patterns for 2020-2050, and analyze the key drivers of these water supply-demand changes. finally quantify water supply-demand risks under the land-climate combination patterns, identify the main factors influencing these water supply-demand risks, and propose management and policy recommendations aligned with regional development. The framework and workflow of this research approach are illustrated in Fig. 2.



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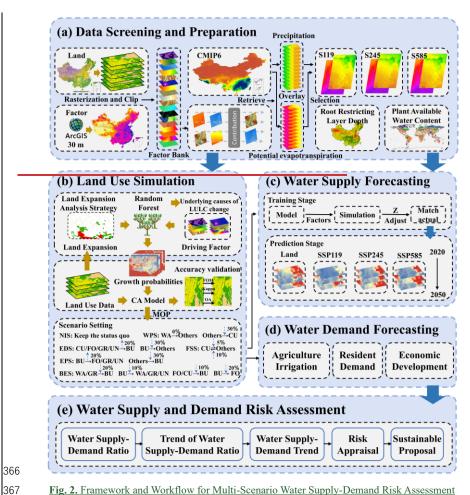


Fig. 2. Framework and Workflow for Multi-Scenario Water Supply-Demand Risk Assessment

2.3.1 Land-Climate Model Setting

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To explore the diverse possibilities for TRB's development, this study integrated the "Aksu Prefecture National Economic and Social Development 14th Five-Year Plan and Long-Range Objectives Through the Year 2035", the "Aksu Prefecture National Economic and Social Development Statistical Bulletin (2020-2024)", the "Aksu Prefecture Territorial Spatial Plan (2021-2035)", the "Xinjiang Uygur Autonomous Region Territorial Spatial Plan (2021-2035)", and previous research findings (Kulaixi

et al., 2023; Song et al., 2025) to establish six land development scenarios.

Natural Increase Scenario (NIS): Based on the land evolution process in the TRB from 2000 to 2020, this scenario maintains the current land transition processes, adds no new policy influences, and imposes no restrictions on the transfer probabilities between land use types. It serves as a baseline and reference for the other scenarios. It also functions as a control for observing transitions in the other restricted scenarios.

Food Security Scenario (FSS): Based on the characteristics of the TRB region, this scenario emphasizes food security and enhances agricultural productivity. It reduces (by 5%) the transfer probability of cultivated land to other land use types while increasing (by 10%) the transfer probability from other land use types to cultivated land.

Economic Development Scenario (EDS): Driven by accelerating urbanization and economic development needs, this scenario enhances economic construction and fundamental urban capacity. It increases (by 20%) the transfer probability from cultivated land, forest land, grassland, and unused land to built-up land, keeps the transfer probability from water bodies to built-up land unchanged, and simultaneously protects the TRB's economic infrastructure by reducing (by 30%) the probability of built-up land converting to other land use types except cultivated land.

Water Protection Scenario (WPS): Addressing water scarcity and the need for aquatic ecological balance, this scenario prioritizes safeguarding ecological functions such as water resource protection and water conservation from infringement. It prohibits the encroachment of existing water body areas by other land use types and reduces (by 30%) the transfer probability from other land types to cultivated land.

Ecological Protection Scenario (EPS): Given the ecological fragility and sensitivity of the TRB, this scenario aims to enhance the resilience of its eco-environment. It restricts (by 30%) the transfer probability from other land use types to built-up land and increases (by 20%) the transfer probability from built-up land to forest land, grassland, water bodies, and unused land.

Balanced Economy and Ecology Scenario (BES): Responding to the dual demands of economic development and ecological governance in the TRB, this This scenario seeks parallel development of urbanization and ecological conservation. It reduces (by

20%) the transfer probability from grassland and water bodies to built-up land, and reduces (by 10%) the transfer probability from cultivated land and forest land to built-up land. Building upon this, it reduces (by 20%) the transfer probability from built-up land to forest land, and reduces (by 10%) the transfer probability from built-up land to water bodies, grassland, and unused land.

In response to the increasingly severe climate change, combining historical rainfall and potential evapotranspiration trends in the TRB (Fig. 1). this scenarios with Shared Socio-economic Pathways (SSP) and Representative Concentration Pathways (RCP) under CMIP6 were selected. While SSP describes possible future socio-economic developments, RCP depicts future greenhouse gas concentration and radiative forcing scenarios (O'Nill et al., 2016, 2017). Here, the typical SSP-RCP scenarios from the second-generation climate model (MRI-ESM2.0) as developed by the Meteorological Research Institute (MRI) of Japan were used. This includes: i) land, to compare current and future climate change; ii) SSP119, the lowest radiative forcing scenario with radiative forcing of \approx 1.9 W/m² by 2100; iii) SSP245, a medium radiative forcing scenario that stabilizes at \approx 4.5 W/m² by 2100; iv) SSP585, a high forcing scenario with emissions rising to 8.5 W/m² by 2100.

2.3.2 Land Use Projections

This study employed the PLUS model to predict land use evolution trends in the TRB. The PLUS model consists of the Land Expansion Analysis Strategy (LEAS) and the CA based on multi-type random patch seeds (CARS) (Liang et al., 2021). The LEAS module utilizes the random forest algorithm to explore the relationships between multiple driving factors and different land types, thereby determining the development potential for each land use type (Shi et al., 2023). The CARS module simulates patches of different land types by integrating a transition matrix and neighborhood weights of land use types to achieve the prediction outcome. In this study, the sampling rate of the random forest was adjusted to 0.2 and the number of decision trees was set to 60 to adapt to the geographical environment of the TRB. —

Fig. 3. Observed (2015/2020) and simulated land use maps of the Tailan River Basin (left/middle-

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We selected the Figure of Merit (FOM), Overall Accuracy (OA), and Kappa index (Liu, et al., 2017) to measure the accuracy of the simulations. To enhance the applicability and precision of the PLUS model, the collected 19 driving factors were used as a 'factor bank'. Under consistent other simulation parameters, factors with lower contribution capabilities were systematically removed, and land use patterns for both 2015 and 2020 were simulated. Driving factors were screened based on the random forest algorithm within the LEAS module and the evaluation metrics. When the number of driving factors was reduced to 13, the simulation achieved the highest accuracy (Tab. 3S1) and exhibited strong consistency (Fig. S23). Consequently, this study adopted these 13 driving factors for subsequent simulations.

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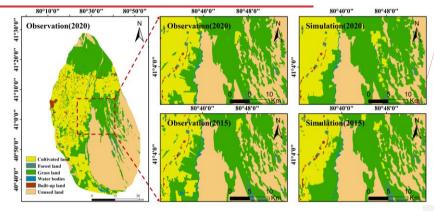


Table 3. Accuracy metrics of the PLUS model during different validation periods.

Simulation time	Number of factors	OA	Kappa	FOM
	7	0.91	0.86	0.17
2015	13	0.92	0.88	0.18
	19	0.91	0.87	0.18
2020	7	0.88	0.83	0.19
2020	13	0.90	0.85	0.25

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2.3.3 Water Supply and Demand Forecasting

2.3.3 Water Supply and Demand Forecasting

(1) Water Supply Forecasting

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This study utilized the water yield module of the InVEST model to predict changes in water yield within the TRB (Tailan River Basin). The Budyko framework (Budyko, et al., 1974) was applied to determine the difference between precipitation and actual evapotranspiration for each grid cell, which was then used to calculate water yield (Chen, et al., 2024a). The calculation formula is as follows:

$$Y_{(x)} = \left(1 - \frac{AET_{(x)}}{P_{(x)}}\right) \times P_{(x)} \tag{1}$$

where $Y_{(x)}$ is the annual water yield of grid cell x; $AET_{(x)}$ is the actual

evapotranspiration in grid cell x; and $P_{(x)}$ is the annual precipitation in grid cell x.

Evapotranspiration of vegetation under the various land use types was calculated (i.e.,

463 $\frac{AET(x)}{P(x)}$) after Zhang et al. (2004) as follows:

$$\frac{AET_{(x)}}{P_{(x)}} = 1 + \frac{AET_{(x)}}{P_{(x)}} - \left[1 + \left(\frac{PET_{(x)}}{P_{(x)}}\right)^{\omega}\right]^{1/\omega}$$
 (2)

where PET_(x) is the potential evapotranspiration (mm) of grid cell x, and ω is an empirical value related to natural climate and soil properties. The term $\omega(x)$ is calculated after Donohue (2012) as:

$$\omega(x) = Z \frac{AWC_{(x)}}{P_{(x)}} + 1.25 \tag{3}$$

$$AWC_{(x)} = min(MaxSoilDepth_{(x)}, RootDepth_{(x)}) \times PAWC_{(x)}$$
 (4)

where Z is a seasonal constant of the water yield model, representing hydrogeological characteristics such as regional precipitation distribution. Based on "the Wensu County Water Resources Development Plan for the 14th Five-Year Plan Period" and "the Comprehensive Report on the Tailan River Basin Planning", the surface water resources volume in the plain area was determined to be 65×10^5 m³. Through manual optimization, it was found that when the model parameter Z = 7.5, the

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discrepancy between the simulated and observed values was minimized (Fig. 4S3).-

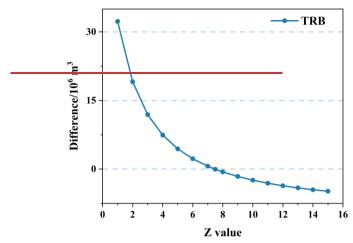


Fig. 4.

Discrepancy between observed and simulated surface water resources volume in the Tailan River-

479 Basin under different Z values.

 AWC_(x) is the effective water content of grid cell x; PAWC_(x) is the effective water content of vegetation in grid cell x; MaxSoilDepth_(x) is the maximum soil depth in grid cell x; and RootDepth_(x) is the root depth in grid cell x. The term PAWC_(x) is as follows (Zhou et al., 2005):

$$PAWC_{(x)} = 54.509 - 0.132SAND_{(x)} - 0.003(SAND_{(x)})^{2} - 0.055SILT_{(x)}$$
$$- 0.006(SILT_{(x)})^{2} - 0.738CLAY_{(x)} + 0.007(CLAY_{(x)})^{2}$$
$$- 2.699OM_{(x)} + 0.501(OM_{(x)})^{2}$$
(5)

where $SAND_{(x)}$, $SILT_{(x)}$, $CLAY_{(x)}$, and $OM_{(x)}$ respectively stand for sand, silt, clay, and organic matter contents of grid cell x.

(2) Water Demand Forecasting

As indicated by "the Wensu County Statistical Bulletin on National Economic and Social Development" and "the Aksu Statistical Bulletin on National Economic and Social Development", the water use structure in the TRB (Tailan River Basin) is well-defined, primarily sourced from agricultural irrigation, residential consumption, and economic development activities. Therefore, this study conducted separate projections for agricultural water demand, domestic water demand, and economic water demand

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within the TRB. In order to account for the impact of climate change on the average crop water requirement in the TRB, and based on the findings of Li et al. (2020), which indicate that a temperature increase of 2 °C leads to an increase in the average crop water requirement of 19 mm, the formula for calculating irrigation water demand per hectare was derived as follows:

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$$\Delta cwd_{(a,b)} = 9.5 \times (T_2 - T_1) \tag{6}$$

$$n_{(a,b)} = d_0 + \Delta cw d_{(a,b)} \tag{7}$$

where Δ cwd $_{(a,b)}$ represents the change in average crop water requirement for grid cell b in year a, T1 denotes the air temperature for a grid cell during the baseline period, and T2 denotes the air temperature for the same grid cell during the change period. n(a,b) is the irrigation water demand per hectare for grid cell b in year a under climate change impacts, and do is the irrigation water demand per hectare during the baseline period. Therefore, the calculation formulas for agricultural water demand, domestic water demand, and economic water demand are as follows:

$$pop_{(a,b)} = \frac{p_{(a,b)}}{\sum_{b=1}^{n} p_{(a,b)}} \times POP_{a}$$

$$gdp_{(a,b)} = \frac{g_{(a,b)}}{\sum_{b=1}^{n} g_{(a,b)}} \times GDP_{a}$$
(9)

$$gdp_{(a,b)} = \frac{g_{(a,b)}}{\sum_{b=1}^{n} g_{(a,b)}} \times GDP_a$$
 (9)

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$$WD_{(a,b)} = pop_{(a,b)} \times l_{(a,b)} + gdp_{(a,b)} \times m_{a(a,b)} + agr_{(a,b)} \times n_{(a,b)}$$
 (10)

where $p_{(a,b)}$ and $g_{(a,b)}$ are respectively the initial population and economic status of grid cell b in year a; POPa and GDPa are respectively the population and GDP in year a; $pop_{(a,b)}$ and $gdp_{(a,b)}$ respectively the calibrated population and GDP of grid cell b in year a; and $agr_{(a,b)}$ is the cultivated land area of grid cell b in year a. the The terms l_a , ma, and na respectively represent the per capita water use, water use per 10,000 Yuan of GDP, and irrigation water use per hectare of farmland in year a. To exclude recharge from the mountain in the study area, the amount of surface water resources in the mountains was equally dispersed in a raster. The population and GDP for 2030-2050 were determined using linear regression method. To exclude the water contribution from the upper reaches of the Tailan River to this study, the multi-year average runoff from the upper reaches was evenly allocated to each grid cell to reduce its influence on water demand calculations. Additionally, this study employed linear regression to

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project the population and GDP for the period 2030–2050, which was used to support the prediction of the temporal change in water demand within the TRB from 2030 to 2050.

2.3.4 Risk Framing of Water Supply and Demand

 The water supply-demand risk framework serves as a crucial tool for assessing regional water supply-demand risks. Moran_—(2017—) classified the computational results generated within this framework into seven categories (Tab. 42), enabling the assessment of regional water risk levels by calculating the water supply-demand relationship and facilitating the quantification of regional water supply-demand risk grades. This framework comprises four indicators: the water supply-demand ratio, the trend in the water supply-demand ratio, the water supply trend, and the water demand trend. The calculation procedures for these indicators are as follows:

1) The water supply and demand ratio that expresses spatial heterogeneity of water supply and demand contradictions:

$$R_{(x)} = WY_{(x)} / WD_{(x)}$$
(9)

where $R_{(x)}$ is the water supply-demand ratio of grid cell x; and $WY_{(x)}$ and $WD_{(x)}$ are respectively the water supply and demand of grid cell x.

2) The trend of water supply-demand ratio expresses the relative changes in water supply and demand:

$$R_{tr} = R_i - R_i \tag{10}$$

where R_{tr} is the difference between water supply-demand ratios in years i and j; R_{i} and R_{j} are respectively the water supply-demand ratios in years i and j.

3) The trend of water supply and demand volume expresses the absolute changes in water supply and demand volume:

$$S_{tr} = WY_i - WY_i \tag{11}$$

$$D_{tr} = WD_i - WD_i \tag{12}$$

where S_{tr} and D_{tr} are respectively the differences in water supply and demand-volumes; WY_i and WY_j respectively the water supply volumes in years i and j; WD_i and WD_j respectively the water demand volumes in years i and j.

Table 42. Assessment of water supply and demand risk level in the study area

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Grade code	Risk grade	Water supply-demand ratio (R)	Water trend of supply–demand ratio (R_{tr})	Trend of water Supply (S_{tr}) and demand (D_{tr})
I	Extinct/ Dormant	R = 0	$R_{tr} < 0$	_
II	Critically endangered	0 < R < 1	$R_{tr} < 0$	$S_{tr}\!<\!0,D_{tr}\!\ge\!0$
III	Endangered	0 < R < 1	$R_{tr} \! \geq \! 0$	$\begin{split} S_{tr} &< 0, D_{tr} < 0 \text{ or} \\ S_{tr} &\geq 0, D_{tr} \geq 0 \end{split}$
IV	Dangerous	0 < R < 1	$R_{tr} \! \geq \! 0$	$\begin{split} S_{tr} &< 0, D_{tr} < 0 \text{ or} \\ S_{tr} &\geq 0, D_{tr} \geq 0 \end{split}$
V	Undersupplied	0 < R < 1	$R_{tr} < 0$	$S_{tr} \geq 0, D_{tr} < 0$
VI	Vulnerable	$R \ge 1$	$R_{tr} \ge 0$	_
VII	Safe	$R \ge 1$	$R_{tr} \geq 0$	_

3 Results

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3.1 Spatial heterogeneity of land use to multi scenario

The evolution of land use in the TRB from 2020 to 2050 under six scenarios was simulated using the PLUS model. Overall, the land use structure remained relatively stable across the multiple scenarios, with the most significant changes primarily manifested in cultivated land (33%) and grassland (29%) areas (Fig. 53). Notably, grassland area generally exhibited significant degradation (with an average reduction of 535.36 km²), whereas cultivated land area expanded substantially (The contribution of population is the highest (0.22) (Fig. 4)) due to factors such as policy incentives and population growth (with an average increase of 524.87 km²). Under the NIS, the intensity of cultivated land reclamation continuously increased, with its proportion jumping from 33% (2020) to 46% (2050). A significant portion of this expansion stemmed from the reclamation of grassland. Simultaneously, the encroachment of builtup land also constituted a major component of grassland conversion. Compared to NIS, the FSS resulted in a greater expansion of cultivated land (545.28 km²). This scenario emphasizes intensive land use and promotes sustainable cultivated land development through the consolidation of fragmented farmland. The cultivated land expansion under FSS primarily originated from the conversion of grassland. Under the EDS, the area of built-up land surged from 62.88 km² (2020) to 113.05 km² (2050), significantly exceeding that in other scenarios. Urban expansion primarily encroached upon

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eultivated land and grassland (Liu et al., 2015). Relative to NIS, the WPS mitigated grassland reclamation and degradation, increased water conservation and ecological land, and augmented grassland area through soil conservation measures and the development of wasteland. Building upon WPS, the EPS further restricted human activities, resulting in the smallest built-up land area (104.08 km²). While controlling the growth rate of cultivated land area, it significantly increased the area of ecological land, such as grassland and water bodies, thereby further restoring the fragile ecosystems in the arid region. As a key measure to balance ecology and economy in the arid oasis region, the BES maintained a relatively high cultivated land area (531.20 km²) to safeguard the agricultural economic backbone. Simultaneously, it ensured that ecological land, such as woodland and water bodies, remained free from encroachment. Furthermore, it involved further development of unused land (wasteland and saline-alkali land), converting it into grassland (31.08 km²) with ecological conservation functions.

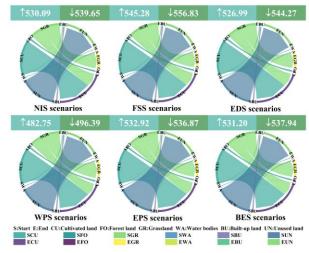
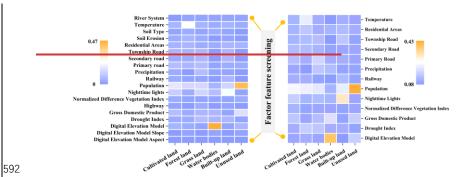


Fig. 5-3 Transfer process under six land-use scenarios in the Tailan River Basin, 2020–2050 (Scenario labels indicate cultivated land expansion area (blue; in km²) and grassland degradation area (green; in km²)).



Different land types exhibit significantly varying degrees of responsiveness to driving factors due to differences in their spatial demand and evolutionary trajectories (Fig. 64). Specifically, population plays a core driving role in the evolution of multiple land types: it exhibits the highest contribution rates to cultivated land (0.2022), forest land (0.19), grassland (0.17), built-up land (0.18), and unutilized land (0.43). Other key driving factors also show specific influences: the Nighttime Light Index has relatively high contributions to cultivated land (0.12) and built-up land (0.29), the Aridity Index to forest land (0.11) and grassland (0.09), and the Digital Elevation Model (DEM) also contributes significantly to water bodies (0.38).

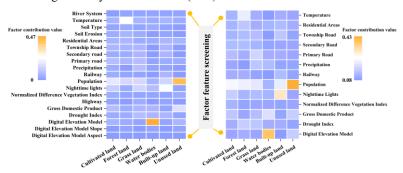


Fig. 64. Driver banks (left) and screened high contributors (right)

3.2 Multi-Scenario Water Supply-Demand Dynamics

(1) Variation in water supply

 Based on the InVEST model, the variation trends of water supply under different climate change and land use scenarios were investigated. The spatial distribution of water resources supply remains consistent across scenarios, with a stable water supply

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pattern (Fig. 7a5a). This pattern demonstrates significantly higher water supply in the northern region than elsewhere, which is closely linked to the spatial distribution of precipitation in the TRB. During 2020-2050, water supply trends under different scenarios show distinct variations: both Land and S245 exhibit an upward trend, with S245 increasing at a significantly faster rate than land. In contrast, the water yield capacity of S119 and S585 gradually declines over time, though their decreasing trends differ substantially (Tab. 582). Furthermore, the contribution of water yield capacity from different land types to water supply also varies, with grassland providing significantly higher water supply than cultivated land (Fig. 5a). Using the scenario maintaining current rainfall and potential evapotranspiration (Land) as the baseline, TRB's water supply fluctuates under different land use scenarios, ranging from 64.78 × 10^5 m³ to 65.7×10^5 m³. Under different climate scenarios, TRB's water supply shows pronounced variations, with a fluctuation range of 25.33 × 10⁵ m³ to 162.2 × 10⁵ m³ when referenced against the NIS baseline scenario. The highest water supply in TRB $(162.8 \times 10^5 \text{ m}^3)$ occurs under the S245-FSS, while the lowest $(25.23 \times 10^5 \text{ m}^3)$ is observed under the S119-EPS._

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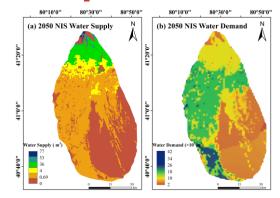


Fig. 5. Spatial patterns of water supply (a) and water demand (b) in the Tailan River Basin under the Natural Increase Scenario (NIS) in 2050.

Fig. 7. Spatial patterns of water supply (a) and water demand (b) in the Tailan River Basin for 2050

Table 5. Dynamics of Water Supply and Demand in the Tailan River Basin Under 24 Land Climate

Combination Scenarios (2020–2050)

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(2) Variation in water demand

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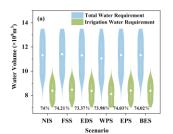
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Compared with water supply, the spatial distribution and pattern of water demand also remain relatively consistent and stable across different scenarios (Fig. 765b) This pattern exhibits stronger water demand capacity in the southwestern and central-eastern regions but weaker capacity in the northern and southeastern areas, which is closely associated with the spatial distribution of land use and population aggregation density in the TRB (Fig. 5b). During 2020-2050, water demand under all scenarios shows a continuous upward trend, though with significant variations in the rate of increase. Furthermore, the contribution of water demand capacity from different land types varies markedly, with cultivated land and built-up land demonstrating stronger demand capacity, while unutilized land shows the weakest capacity. Using the scenario maintaining current rainfall and temperature (Land) as the baseline,. TRB's water demand exhibits significant variations under different land use scenarios (Tab. <u>5S2</u>), ranging from 1575 × 10⁵ m³ to 4935 × 10⁵ m³. Under different climate scenarios, TRB's water demand displays similar upward trends over time, with a fluctuation range of 1887×10^5 m³ to 5316×10^5 m³ relative to the NIS baseline scenario. The highest water demand (5390 × 10⁵ m³) occurs under the S585-FSS scenario, whereas the lowest (1575 × 10⁵ m³) is observed in the Land-WPS scenario. Agricultural water use has consistently constituted the primary consumption component in the TRB. Across all land and climate change scenarios, irrigation accounts for over 70% of the total share (Fig. 8a6a, b). Although the proportion of irrigation water gradually decreases over time, its total volume continues to increase (Fig. 8e6c). Nevertheless, unilateral studies of water supply or demand alone cannot directly reflect water resource allocation capacity. The impacts arising from supply-demand imbalances remain unclear and warrant further investigation. To better elucidate the impacts of water supply-demand dynamics on TRB's water resources, in-depth analysis of regional water security risks is required, which will facilitate the formulation of tailored water management and conservation strategies.



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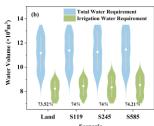
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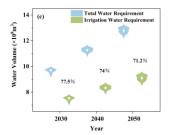


Fig. 86. Dynamics of total water demand and agricultural irrigation demand with

proportional distribution across latitudinal gradients in the Tailan River Basin, 2030–2050(a)

Different land change scenarios (Natural Increase Scenario (NIS)/ Food Security Scenario (FSS)/

Economic Development Scenario (EDS)/ Water Protection Scenario (WPS)/ Ecological Protection

Scenario (EPS)/ Balanced Economy and Ecology Scenario (BES)); (b) Different climate change

scenarios (Land/S119/S245/S585); (c) Temporal evolution (percentages in the figure represent the

proportion of irrigation water to the total water demand),

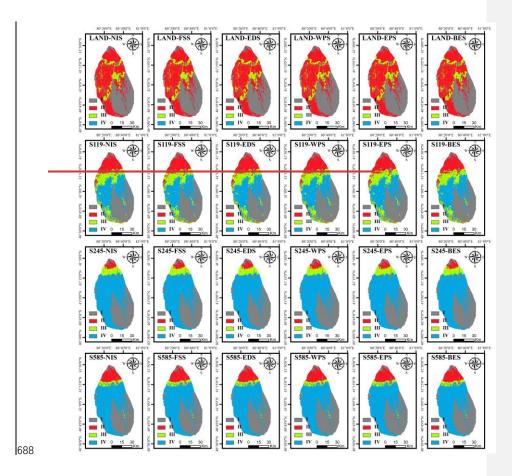
3.3 Multi-Scenario Water Supply-Demand Risks and Attribution

To assess water supply-demand risks in the TRB region, an evaluation framework was established using four indicators: water supply-demand ratio, trend of water supply-demand ratio, water supply trend, and water demand trend. Spatial patterns of water supply-demand risk in the TRB exhibit heterogeneity across scenarios (Fig. 97). Under identical climate change scenarios, risk variations across land use scenarios are insignificant, with consistent spatial distribution of risk classification levels. This consistency arises from unchanged rainfall and potential evapotranspiration patterns. Although risk classification levels vary under different climate scenarios, no grid cell in the TRB escapes hazardous (Level IV) risk (Tab. 3-2 indicates a 7-level classification system). This is closely linked to continuously increasing water demand in the TRB. Using NIS as the baseline, the scenario maintaining current rainfall and potential evapotranspiration (Land) shows the most severe water scarcity: Level II risk accounts for 51.31%, while Level IV risk constitutes merely 0.85%. Under the other three climate scenarios, water supply-demand risks are alleviated, with Level IV risk proportions being 29.24% (S119), 53.60% (S245), and 49.34% (S585) respectively (Fig. 108). The S245-EPS scenario achieves maximum risk mitigation in the TRB, as its rainfall levels

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increase steadily per decade among the three climate scenarios (Fig. 1d), thereby alleviating regional water stress. While the TRB's harsh current climate exacerbates water risks, future climatic changes may moderately alleviate these risks compared to present conditions. In summary, by 2050 the entire TRB will face water supply-demand crises, with at least 46% of the area subjected to endangered (Level III) risk, including no less than 10% of land confronting critical endangered (Level II).



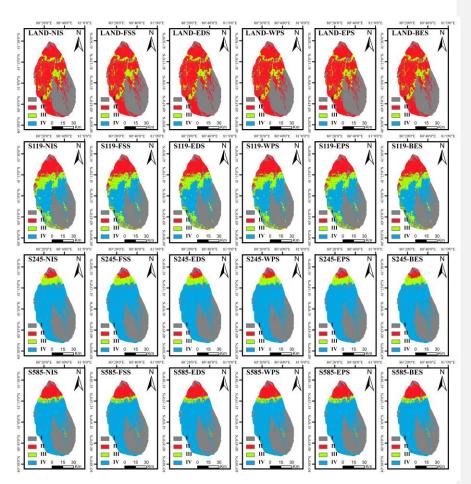


Fig. 97, Spatial evolution of water supply-demand risk classification levels in the Tailan River

Basin (TRB) under 24 climate-land combination scenarios (2020–2050)

(Land change scenarios (Natural Increase Scenario (NIS)/ Food Security Scenario (FSS)/

Economic Development Scenario (EDS)/ Water Protection Scenario (WPS)/ Ecological Protection

Scenario (EPS)/ Balanced Economy and Ecology Scenario (BES)); Climate change scenarios

(Land/S119/S245/S585): (Color gradient indicates decreasing risk from Level I (highest) to Level

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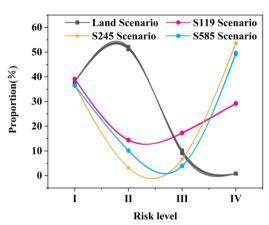


Fig. 108. Temporal variation in proportional distribution of risk classification levels across the Tailan River Basin (TRB) under 24 climate-land combination scenarios (2020–2050)

(decreasing risk from I to VII)

4 Discussion

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4.1 Multi-Scenario Land Use Spatial Patterns

Land use changes alter regional hydrological processes and water resource patterns—such as infiltration, groundwater recharge, baseflow, and runoff—thereby affecting regional water supply and demand dynamics (Lin et al., 2007). During 2020-2050, land type transitions in the TRB will predominantly involve cultivated land and grassland, which will profoundly influence water supply-demand dynamics (Fig. 3). In this study, the Food Security Scenario (FSS) prioritizes cohesive cultivated land expansion with high intensification and contiguity (1,937.58 km²). Although rapid cultivated land growth directly boosts regional agricultural economies, it significantly increases agricultural irrigation water demands (Sharofiddinov et al., 2024), intensifying pressure on water supply-demand balance. Notably, under the Economic Development Scenario (EDS), further intensification of human activities exacerbates this pressure. Rapid urbanization not only elevates domestic and industrial water demands (He et al., 2021) but may also degrade water quality and availability by altering surface runoff and amplifying non-point source pollution (Strokal et al., 2021). In contrast, the Ecological Protection Scenario (EPS) reinforces ecological barriers by restraining agricultural expansion and limiting grassland conversion, thereby reducing 设置了格式: 字体颜色: 红色

water consumption. The Water Protection Scenario (WPS) mitigates land fragmentation, potentially preserving more natural hydrological processes and water conservation functions. Together, these measures slow ecological degradation and indirectly support long-term water sustainability. These results demonstrate that the rate of natural resource consumption by human activities (particularly agricultural practices) far exceeds the rate of natural recovery, and this antagonistic relationship weakens as human interventions intensify. Consequently, the Balanced Economic and Ecological Development Scenario (BES) seeks to reconcile economic growth with ecological conservation (especially water resources) by moderately controlling cultivated land and construction land expansion (reducing encroachment on grassland to 689.17 km²), making it a prioritized land use model for the future. Different land use scenarios highlight the critical leverage of land use policies in water resource management. The significant spatial heterogeneity across the region calls for targeted strategies to alleviate future pressures from human activities (especially agriculture) on water resource systems. The selection of land use patterns determines the spatial layout and functions of urban systems (Zhao et al., 2021), serves as a fundamental element for territorial spatial planning and configuration (Chen et al., 2023b), and constitutes a critical approach to optimizing regional spatial structure, coordinating development processes, and achieving ecologically-economically synergistic sustainability within limited resources and space (Feng et al., 2025; Wu et al., 2025). In this study, the FSS adopts a more cohesive evolution trajectory for cultivated land (Fig. 5), exhibiting high intensification and contiguity (1937.58 km²). As an important economic driver in the TRB, the sustained rapid expansion of cultivated land will directly stimulate regional agricultural development, indicating that human activities determine land use trajectories and patterns. However, due to challenges in saline alkali land remediation, cultivated land expansion encroaches on additional grassland, which will induce severe ecological degradation. The EDS further amplifies human impacts on land use, exacerbating compression for grassland (682.84 km²). Chen et al. (2023) demonstrates that ecological degradation from urbanization parallels that from cultivated land expansion, while more intensive human activities in urbanized areas may inflict greater

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ecological damage. This implies that under anthropogenic pressure, heightened agricultural intensification and settlement land demand eneroach upon more ecological land (forest and grassland), generating increased ecological deficits. The WPS strengthens ecological barriers by curbing agricultural land expansion and restricting grassland conversion, thereby reducing water resource consumption. Building on this, the EPS mitigates land fragmentation by controlling built-up land growth rates. Results indicate both WPS and EPS decelerate grassland degradation. This reveals that anthropogenic resource consumption rates substantially exceed natural recovery rates, with their antagonistic relationship weakening as human activities intensify. To maintain antagonistic equilibrium, the BES appropriately reduces cultivated land and built-up land areas, alleviating human encroachment on grassland (689.17 km²). It represents a prioritized model for future decision-making. Diverse land use scenarios reflect regional policy orientations and latent conflicts (Qi et al., 2025). Significant spatial heterogeneity necessitates targeted land management strategies to mitigate future anthropogenic impacts on natural ecosystems.

Table S1, indicates that using more driving factors does not necessarily improve model performance, and the selection of these factors is a critical source of uncertainty in the results. Although the Random Forest algorithm effectively addresses multicollinearity among factors, complex interactions between driving factors can still introduce noise and increase the predictive uncertainty of simulations (Liang et al., 2021). Specifically, this study achieved optimal simulation accuracy with 13 driving factors. Adding factors with low contributions beyond this number distorted the direction and quantity of simulations, thereby reducing model accuracy. Conversely, when the number of factors was reduced to 7, simulation performance declined significantly. This high sensitivity suggests that the PLUS model is vulnerable to input uncertainty; the absence of key driving factors directly increases bias in the accuracy of simulations. The significant differences in factor contributions (Fig. 4) further highlight uncertainties arising from human activities. The intensity of human activities and population distribution density profoundly influence land use change processes. Changes in population size trigger cascading effects on land resources, agricultural

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ecosystems, and water resources by altering wealth levels and food calorie demands (Beltran et al., 2020; Harifidy et al., 2024). Land use change is a multidirectional process, and quantifying human activities (e.g., economic dynamics and population migration) remains challenging. Climate variability further exacerbates simulation uncertainties. Therefore, it is essential to employ multi-scenario simulations to provide decision-makers with a range of possible future land change pathways, thereby reducing policy risks. Results from Table 3 reveal that increasing the number of driving factors does not necessarily improve outcomes. Although random forests can effectively handle multicollinearity among factors, complex correlations and interactions between driving factors may still compromise simulation accuracy (Liang et al., 2021). Specifically, utilizing 13 core driving factors yields optimal simulation performance. Adding factors with low contribution rates beyond this threshold disturbs patch simulation directionality and quantity, thereby reducing precision. Conversely, reducing the number to 7 significantly degrades simulation quality, indicating high sensitivity of the PLUS model to feature factors. The absence of key drivers directly impairs patch simulation accuracy. Significant disparities in factor contribution rates (Fig. 6) demonstrate that human activity intensity and population distribution density govern land evolution processes. Shifts in population size trigger changes in affluence levels and food calorie demand, generating cascading effects on land resources, agroecosystems, and water resources (Beltran et al., 2020; Harifidy et al., 2024).

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4.2 Land Use and Climate Change Impacts on Water Supply-Demand Dynamics

(1) Impacts of Climate Change and Land Use on Water Supply

Water supply in the TRB is jointly constrained by human activities and climate change. Under the same climate change conditions, there are differences in water supply between different land use scenarios (Tab. S2), and these differences are caused by different land use structures (Jia et al., 2022). Analysis of variance across the 24 climate-land combination scenarios revealed that their variability range was significantly lower than that of different climate change scenarios (Tab. S2), indicating that climate change (precipitation) exerts a more pronounced influence on water supply

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(2) Impacts of Climate Change and Land Use on Water Demand

Water demand in the TRB is also constrained by both human activities and climate

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change (Tab. S2). Under the same land use scenario, water demand varies across different climate scenarios, with this variation driven by temperature-induced changes in irrigation water use (Li et al., 2020). Analysis of variance across the 24 climate-land combination scenarios revealed that the impact of climate change on water demand was significantly lower than that of land use changes (Tab. S2), indicating that human activities (land use) exert a more substantial influence on water demand in the TRB than climate change. However, climate change impacts on water demand are substantially lower than those from land use changes (Table 5), indicating that human activities (land use) exert a more significant influence on TRB water demand than climate change. It is clear that irrigation water consumption accounts for the majority of TRB water consumption (Fig. 86). Changes in TRB's irrigation water are closely linked to (1) conversions between cultivated land and other land types, and (2) adjustments in planting patterns within cultivated areas. Studies demonstrate that volatile land allocation significantly affects agricultural irrigation, particularly through land type conversions (Cao et al., 2024). Simultaneously, land fragmentation levels influence water user numbers, while changes in irrigated area and frequency intensify irrigation water pressure (Sharofiddinov et al., 2024). This indicates that expanding cultivated land areas drive increased irrigation water usage (Liu et al., 2025a), aligning with our findings. Additionally, planting area and planting structure significantly impact irrigation water use (Chen et al., 2020). Sun et al. (2024) confined irrigation water within manageable levels while boosting yield and carbon sequestration by adjusting rice, maize, and soybean cultivation areas; Other research reduced irrigation water by 34.48% while decreasing crop greenhouse gas emissions by 10% through planting structure optimization (Li et al., 2025a). Moreover, interactions between irrigation technology and planting structure adjustments affect irrigation demand. Wu et al. (2024) found that combining deficit irrigation with high-density planting reduces irrigation water by 20% without compromising cotton yield. Furthermore, growers' strong traditional agricultural values make produce value and labor costs more critical concerns than irrigation water consumption (McArthur et al., 2017; Nourou et al.,

2025). For instance, widespread maize cultivation (an economic crop) in the TRB

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substantially increases regional irrigation water volumes (Huang et al., 2015). Consequently, adjusting regional cropping structures within a macro-agricultural framework is crucial for ensuring sustainable water use and safeguarding growers' economic returns.

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Currently, most studies on water supply and demand focus primarily on the unilateral impacts of either climate or human activities (land use changes) (Wen et al., 2025; Bai et al., 2025; Deng et al., 2024), or emphasize recent temporal changes. For example, Chen et al. (2024a) quantitatively evaluated the water conservation function of the Yangtze River over the past 40 years, while Ma et al. (2023) analyzed the effects of land use and land cover (LULC) changes on water yield (WY) in the Bosten Lake region from 2000 to 2020. However, studies have shown that complex interactive feedback mechanisms exist between climate change and land use, but their degrees of influence on water resources differ (Qi et al., 2025). Changes in water supply and demand are also affected by their combined impacts (Dey et al., 2017; Tan et al., 2025; Tian et al., 2025). Therefore, it is essential to assess future water supply-demand relationships under the dual influences of climate and land use changes. Based on comprehensive calculations across 24 climate-land combination scenarios, this study revealed a notable disparity between the change in water supply (137.47×10⁵ m³) and the change in water demand (3815×10⁵ m³), indicating that human activities have a greater impact on water resources in the TRB than climate change. This significant imbalance between water supply and demand will have profound implications for regional water supply-demand risks. Climate change and human activities exert nonsingular impacts on water resources (Tan et al., 2025). Concurrently, variations in water supply and demand are jointly influenced by both factors (Tian et al., 2025). Precipitation affects the Normalized Difference Vegetation Index (NDVI) and Net Primary Productivity (NPP), subsequently altering crop phenology and water requirements, which induces changes in underlying surface land types (grassland, forest, and cultivated land). This cascade alters watershed runoff generation/concentration and evapotranspiration, ultimately feeding back to precipitation. This demonstrates complex interactive feedback mechanisms between climate change and land use (Qi et

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al., 2025). Under these mechanisms, distinct climate scenarios yield significantly different water supplies. However, the magnitude of water supply variation (137.47 × 10^5 m³) remains substantially smaller than water demand variation (3815 × 10^5 m³). This indicates that human activities exert a far greater influence on water resources in the TRB than climate change.

4.3 Land Use and Climate Change Impacts on Water Supply-Demand Risks

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By mid-century, water resource vulnerability in the TRB will be profoundly impacted by climate change and human activities (Fig. 7). Global parallels exist: Lu et al. (2024b) demonstrated under multiple land-climate scenarios that synergies between crop production and water yield requirements increase agricultural output but exacerbate water deficits. Chen et al. (2023a) documented significant oasis expansion in China (1987-2017), where increased precipitation and runoff provided partial compensation, yet climate-land changes substantially altered regional water supply. Gaines et al. (2023) found forest cover crucial for maintaining consistent surface water areas across climate-land cover scenarios. Based on this, the study established a water supply-demand risk assessment framework, confirming that water demand continues to increase over time, primarily driven by expanding cropland leading to rising irrigation water needs—a finding consistent with previous reports (Qi et al., 2025). Furthermore, this growing demand will exacerbate water supply-demand risks. These findings confirm that climate change and human activities jointly govern water supply and demand. In our risk assessment framework, water demand trends consistently register negative values (<0), indicating persistent demand growth. This stems from cultivated land expansion driving rising irrigation needs, aligning with prior reports (Qi et al., 2025). Although agriculture water demand share of total water demand declines during 2020-2050, it remains dominant (70%) (Fig. 86). This correlates directly with Section 4.2 findings: (1) conversions between cultivated land and other land types, and (2) adjustments in planting patterns within cultivated areas. Crucially, the arid TRB's limited rainfall cannot meet growing irrigation demands, elevating water supplydemand risk (Land scenario). Compared to current rainfall, three other climate scenarios increase precipitation (2020-2050), improving supply and moderately

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reducing water supply-demand risk (Fig. 97-108). Nevertheless, supply-demand gaps persist at nearly two orders of magnitude, with all areas remaining in "hazardous (Level IV) risk". Thus, human activities remain the primary driver of TRB's water supply-demand risks. Human activities dominate multiple dimensions within these climate-human interactions. Wang et al. (2025) identified human withdrawals as the key driver of reduced runoff and dampened seasonal variability in the Wei River Basin. Similarly, human exceed climate effects on soil moisture decline in China's monsoon loess critical zone (Wang et al., 2024a). Therefore, amid increasing uncertainty, integrating multimethod approaches within water risk frameworks to decipher land-eco-hydrological feedbacks, quantify risks, implement preemptive water regulations, and minimize secondary disasters to ecosystems and societies is imperative.

4.4 Limitations and recommendations for future research

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Nevertheless, this study has certain limitations. (1) While land use change served as the starting point of our research, and multiple driving factors were incorporated for land change simulation, uncertainties in the direction and process of land evolution persist—despite the use of defined transition probability ranges and multiple land use scenarios. These uncertainties may constrain an in-depth exploration of the conversion and evolution processes among different land classes. Although we filtered out driving factors with low contributions, the influence of the TRB's unique geographical environment and ecological processes on land class conversion warrants further investigation. Future studies could explore the impact of arid region ecology-climateenvironment on land use transition processes under the premise of quantifying national land planning and government policy guidance. (2) This study utilized the InVEST model to simulate water yield and employed 24 climate-land combination scenarios to reduce uncertainty in TRB's development. Although the biophysical table in the water yield module was constructed based on the FAO's Crop Evapotranspiration: Guidelines for Computing Crop Water Requirements (https://www.fao.org/) and relevant research (Yan et al., 2020), the evapotranspiration coefficients and plant root depth parameters relied heavily on accurate input values. The dynamic nature of crop growth and water 设置了格式: 字体颜色: 着色 1

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isolate the independent effects of climate, soil, and vegetation on water supply, followup studies should incorporate long-term crop observation data and crop models based on clarified regional cropping structures. This would help refine key parameters (e.g., evapotranspiration coefficients) and disentangle the individual contributions of climate, soil, and vegetation to water supply, thereby reducing uncertainties in the simulation process. Nevertheless, this study has several limitations. (1) Using land change as the starting point, we incorporated multiple drivers for land change simulation. Although our scenario design referenced the TRB's future spatial planning, we did not quantify its impact on urban construction. This may constrain in-depth exploration of land type conversions. Despite screening out drivers with low contribution rates, the influence of TRB's unique geographical setting and ecological processes on land conversion warrants further investigation. Future research could quantify territorial planning and government policies to examine eco-climatic environmental impacts on land use transitions in arid regions. (2) We employed the InVEST model for water yield simulation, using 24 climate land scenarios to reduce uncertainties in TRB's development trajectory. The Biophysical Table in the water yield module utilized FAO's "Crop Evapotranspiration: Guidelines for Computing Crop Water Requirements" (https://www.fao.org/) and research findings (Yan et al., 2022). However, over reliance on precise parameters for evapotranspiration coefficients and root depths introduces uncertainties, given the dynamic nature of crop growth and water demand processes. Subsequent studies could integrate long term crop observation data and crop models based on clarified regional planting structures to refine parameters and reduce uncertainties.

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Located in an arid oasis region, water resources constitute the lifeline for human activities and ecosystems in the TRB. However, the arid and rain-scarce climate of TRB has led to a continuous amplification of the impact of human activities on the ecological environment, with irrigation water demand escalating daily (Chen et al., 2023a; Zhu et al., 2025). Concurrently, human survival necessitates improved living standards and economic development, intensifying human-land conflicts. Based on our findings, we recommend adopting the Balanced Economy and Ecology Scenario (BES)

development model in the TRB, implementing diverse water-saving measures (sprinkler irrigation, subsurface drip irrigation, brackish water irrigation) to control water consumption (Han et al., 2022; Liang et al., 2024), thereby expanding cultivable land reserves.

5 Conclusions

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Elucidating the impacts of climate change and human activities on water supplydemand risks is critical. We applied the PLUS model with six land change scenarios to identify suitable land development strategies for the TRB, and coupled it with the InVEST model under 24 climate-land scenarios to simulate dynamic changes in water supply and demand. Based on this, a water supply-demand risk framework was established to quantify TRB's water supply-demand risks during 2020-2050.Results show that the Balanced Economy and Ecology Scenario (BES) land development model promotes agricultural growth while protecting ecological barriers, adding 531.2 km² of cultivated land by 2050. However, this cultivated land expansion creates a water demand deficit (increasing to 4.87×108 m³), while maximum regional supply reaches only 0.16×108 m3, disrupting water balance. Consequently, the entire TRB will face water crises by 2050, with ≥46% of the area subjected to endangered (Level III) risk. Climate change and human activities jointly drive escalating water supply-demand risks. The root cause lies in persistent cultivated land expansion from intensive human activities, increasing irrigation demand and intensifying supply-demand conflicts. Findings emphasize deep integration of multi-method approaches within the risk framework to decipher land-eco-hydrological feedbacks and consider the complex interrelationships between climate, land, and water supply-demand. Deepening understanding of these linkages is vital for developing effective water scarcity mitigation strategies, providing crucial scientific support for policymakers and land managers.

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1017 gratitude to both the editors and reviewers for their efforts and suggestions. 1018 1019 Code/Data availability The data will be made available on request 1020 1021 **Author contribution** 1022 Conceptualization, YY, YKW, WEW, DYC and XTH; Data curation, YY; 1023 Formal analysis, YY, YKW and DYC; Funding acquisition, PAJ and XTH; 1024 Methodology PAJ, YKW and WEW; Project administration, PAJ, DYC and XTH; 1025 Resources, DYC and XTH; Supervision, PAJ, WEW and XTH; Writing - original 1026 draft, YY; Writing – review & editing, DYC and XTH. 1027 1028 **Competing interests** 1029 The authors declare that they have no known competing financial interests or 1030 personal relationships that could have appeared to influence the work reported in this 1031 1032 paper. 1033

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