

# The effect of climate change on the simulated streamflow of six Canadian rivers based on the CanRCM4 regional climate model

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3 **Abstract**

4

5 The effect of climate change on the hydro-climatology, in particular streamflow, of six major  
6 Canadian rivers (Mackenzie, Yukon, Columbia, Fraser, Nelson, and St. Lawrence) is investigated  
7 by analyzing results from the historical and future simulations (RCP 4.5 and 8.5 scenarios)  
8 performed with the Canadian regional climate model (CanRCM4). Streamflow is obtained by  
9 routing runoff using river networks at 0.5° resolution. Of these six rivers, Nelson and St. Lawrence  
10 are the most regulated. As a result, the streamflow at the mouth of these rivers shows very little  
11 seasonality. Additionally, the Great Lakes significantly dampen the seasonality of streamflow for  
12 the St. Lawrence River. Mean annual precipitation (P), evaporation (E), runoff (R), and  
13 temperature increase for all six river basins in both future scenarios considered here, and the  
14 increases are higher for the more fossil fuel-intensive RCP 8.5 scenario. The only exception is the  
15 Nelson River basin for which the simulated runoff increases are extremely small. The hydrological  
16 response of these rivers to climate warming is characterized by their existing climate states. The  
17 northerly Mackenzie and Yukon River basins show a decrease in evaporation ratio (E/P) and an  
18 increase in runoff ratio (R/P) since the increase in precipitation is more than enough to offset the  
19 increase in evaporation associated with increasing temperature. For the southerly Fraser and  
20 Columbia River basins, the E/P ratio increases despite an increase in precipitation, and the R/P  
21 ratio decreases due to an already milder climate in the Pacific north-western region. The  
22 seasonality of simulated monthly streamflow is also more affected for the southerly Fraser and  
23 Columbia Rivers than for the northerly Mackenzie and Yukon Rivers as snow amounts decrease  
24 and snowmelt occurs earlier. The streamflow seasonality for the Mackenzie and Yukon rivers is  
25 still dominated by snowmelt at the end of the century even in the RCP 8.5 scenario. The simulated  
26 streamflow regime for the Fraser and Columbia Rivers shifts from a snow-dominated to a  
27 hybrid/rainfall-dominated regime towards the end of this century in the RCP 8.5 scenario. While  
28 we expect the climate change signal from CanRCM4 to be higher than other climate models,  
29 owing to the higher-than-average climate sensitivity of its parent global climate model, the  
30 results presented here provide a consistent overview of hydrological changes across six major  
31 Canadian river basins in response to a warmer climate.

32

## 33 1. Introduction

34 As the global population and the standard of living increases so does the strain on  
35 freshwater resources. The natural availability of water is determined by the balance between  
36 precipitation (P) and evaporation (E) (this includes both evaporation and transpiration from  
37 plants). When precipitation exceeds evaporation, which is determined primarily by available  
38 energy, the water that does not evaporate or transpire (either at the surface or after infiltration  
39 into the soil) termed runoff (R) is carried by the rivers to the oceans. The seasonality of  
40 precipitation, its partitioning into snow and rainfall, and the seasonality of snowmelt and  
41 evaporation, all of which are determined by the climate in a given catchment or river basin  
42 eventually determine the seasonality of runoff. As anthropogenic climate change progresses,  
43 changes in the mean annual amounts and the seasonality of these different water budget  
44 components will lead to corresponding changes in runoff (Trenberth et al., 2007). Changes in  
45 precipitation extremes are also expected to lead to corresponding changes in the extremes of  
46 streamflow. The changes in streamflow have implications for floods and power generation. While  
47 runoff is expressed in similar units to precipitation and evaporation (depth of water per unit time,  
48 e.g. mm/s or m/year), streamflow is the volume of water generated per unit time (e.g. m<sup>3</sup>/s or  
49 km<sup>3</sup>/year) and requires multiplication with the area over which runoff is generated. Streamflow  
50 is also routed down the river network which introduces a time lag and attenuation of the peak  
51 runoff.

52 Output from climate and Earth system models (ESMs) remains the primary source of  
53 information for evaluating climate change impacts. Current approaches that rely on information  
54 generated by ESMs, to obtain an estimate of how future streamflow may potentially change, may

55 be classified into two broad categories. The first approach uses simulated runoff directly from  
56 the land surface component of single or multiple climate models which may be routed  
57 downstream to obtain streamflow at the mouths of river basins and at different points along a  
58 given river network (Arora and Boer, 2001; Miller and Russell, 1992; Zhang et al., 2014). Using  
59 direct runoff output from climate models has the benefit that the calculated changes in runoff  
60 are physically consistent with the altered radiative balance of the Earth in response to increases  
61 in the concentrations of greenhouse gases (GHGs). The corresponding changes in the general  
62 circulation of the atmosphere result in the associated changes in near-surface temperature,  
63 precipitation, and the hydrological cycle. However, this approach suffers from three limitations  
64 – 1) the biases in the climate simulated by the climate model, 2) the fact that the land surface  
65 components of climate models are not calibrated for a given river basin but rather designed to  
66 operate in a reasonably realistic way over the whole globe, and 3) the coarse resolution of global  
67 climate models (GCMs). The last limitation is partially addressed when data from finer-resolution  
68 regional climate models is used. The biases in the simulated climate do affect the simulated  
69 runoff for the current climate. Despite this, the approach can effectively capture the effects of  
70 climate change including increased evaporative demand (Winter and Eltahir, 2012), reduced  
71 snowpack (Salathé et al., 2010; Shrestha et al., 2021a), increased winter streamflow, and earlier  
72 snowmelt-driven peak flow (L. Sushama et al., 2006; Poitras et al., 2011). The second approach  
73 attempts to overcome these limitations by downscaling and/or bias-correcting climate from  
74 climate models for future scenarios and uses that to drive a well-calibrated hydrological model  
75 for given catchments or river basins (Gosling et al., 2011; Ismail et al., 2020; Miller et al., 2021;  
76 Yoosefdoost et al., 2022). The second approach is more prevalent for watershed to regional scale

77 impacts and adaptation studies. Given the large effort involved in downscaling and bias-  
78 correcting raw climate data from climate models, most current impact studies use downscaled  
79 and bias-corrected data put together by other groups rather than specifically doing this for their  
80 project. Recent examples include the downscaled and bias-corrected climate data for the  
81 conterminous United States (Thrasher et al., 2013) based on climate model output from the fifth  
82 phase of the Coupled Model Intercomparison Project (CMIP5), and statistically downscaled and  
83 bias-corrected data from five CMIP5 models, available at the global scale, tailored to the  
84 requirements of the Inter-Sectoral Impact Model Intercomparison Project (ISIMIP) (Lange, 2019).  
85 Both these data sets have found large applications in the impacts and adaptation community.  
86 The processes of downscaling and bias correction are distinct, and they both have their inherent  
87 limitations. There are several examples of the limited ability of bias-correction to correct and  
88 downscale variability, and that bias-correction can potentially cause implausible climate change  
89 signals (Maraun, 2016; Maraun et al., 2017). There are also uncertainties, substantial  
90 contradictions, and sensitivity to assumptions between the different downscaling methods  
91 (Hewitson et al., 2014).

92         Finally, while land surface models are typically used within the coupled framework of  
93 climate models, hydrological models are typically used as a standalone model for impact studies.  
94 While the primary output quantities from hydrological models are runoff and streamflow, land  
95 surface models output a range of water, energy, and CO<sub>2</sub> fluxes. The layer of air directly above  
96 the land surface, commonly referred to as the atmospheric or planetary boundary layer, is  
97 affected by surface-atmosphere exchanges of energy and water and extends upward into the  
98 atmosphere. A realistic representation of turbulent fluxes of energy and water in the planetary

99 boundary layer is essential to the transport of moisture and energy through the atmosphere. As  
100 a result, while calibration of hydrological models to reproduce observed streamflow is a routine  
101 exercise (Chegwidden et al., 2019; Hattermann et al., 2018; Huang et al., 2020; Hundecha et al.,  
102 2020), land surface models cannot be calibrated to reproduce a single or a small subset of  
103 quantities. Rather land surface models are expected to reproduce reasonably realistic estimates  
104 of a range of energy, water, and CO<sub>2</sub> fluxes over the whole globe. The philosophy behind land  
105 surface models, as they are used in the context of climate models, is that given 1) a model's  
106 structure and parameterizations, 2) the driving geophysical data for fields such as vegetation  
107 cover, soil depth, and soil texture, and 3) the driving meteorological variables, a model is  
108 expected to reasonably realistically reproduce various components of the water, energy, and  
109 carbon cycle at the global scale. The global scale of land surface models within the framework of  
110 climate models precludes tuning of their parameters for individual grid cells or for a region (e.g.  
111 a river basin) to reproduce a small subset of model outputs.

112 While well-calibrated hydrological models are generally suitable for a given catchment or  
113 a river basin their application cannot be easily extended to large-scale global or regional  
114 hydrologic modelling studies since it is typically not feasible to tune model parameters for all grid  
115 cells in a large domain. For a large region like Canada correctly representing anthropogenic  
116 regulation, using downscaled and bias-corrected climate data from an ensemble of climate  
117 models is a challenging task, and that is why it has only been done for a selected few river basins  
118 in Canada, and that too considering one river basin at a time. In the end, both approaches have  
119 their strengths and limitations for assessing climate change impacts on hydrology and can be  
120 considered complementary to each other.

121 Future hydrologic projections using the second approach (hydrological modes driven by  
122 statistically downscaled and bias-adjusted climate models) are available for selected river basins  
123 in Canada. The results over the Prairies and British Columbia (Shrestha et al., 2021b; Sobie and  
124 Murdock, 2022) generally indicate shorter snow cover duration, earlier snowmelt, and reduced  
125 annual maximum snow water equivalent as the climate warms. Streamflow projections across  
126 Canada generally indicate earlier snowmelt-driven peak flow, increased winter flow, and  
127 decreased summer flow (Budhathoki et al., 2022; Dibike et al., 2021; Islam et al., 2019;  
128 MacDonald et al., 2018; Shrestha et al., 2019). Annual streamflow is projected to increase, with  
129 higher increases in the northern basins (Bonsal et al., 2020; Stadnyk et al., 2021). However, these  
130 projections are based on different climate and hydrological models, downscaling methods,  
131 emissions scenarios, and future periods, and no consistent set of projections is available across  
132 all major river basins of Canada.

133 In this study, we have used the first approach to provide a consistent set of projections  
134 across all major river basins of Canada, while being cognizant of its limitations. We investigate  
135 the effect of climate change on the annual, monthly, and daily streamflow characteristics of six  
136 major Canadian rivers (Mackenzie, Yukon, Columbia, Fraser, Nelson, and St. Lawrence) using  
137 runoff output from simulations performed with version 4 of the Canadian Regional Climate  
138 Model (CanRCM4) (Scinocca et al. 2016). The river basins of the Yukon and Columbia Rivers cover  
139 part of the United States of America as well. We used daily runoff generated from CanRCM4 for  
140 the historical period and for the two future scenarios (representative concentration pathways  
141 (RCP) 4.5 and 8.5). The spatial resolution of runoff data from CanRCM4 is  $0.22^\circ$  which is  
142 equivalent to about 12 km at  $60^\circ$  N (Canada lies between approximately  $42^\circ$ N and  $83^\circ$ N). We

143 then routed this runoff through river networks at 0.5° resolution to evaluate streamflow at the  
144 mouths of major Canadian rivers. The Mackenzie, Yukon, and Fraser Rivers are somewhat less  
145 regulated than the heavily regulated Nelson, Columbia, and St. Lawrence Rivers. The routing  
146 scheme used here does not take into account dams and reservoirs and therefore the modelled  
147 streamflow represents natural streamflow. This aspect is discussed in more detail in Section 2.

## 148 **2. Models and data**

149 Equation (1) summarizes the water balance over a given grid cell or a river basin for a  
150 given timescale.

$$151 \qquad P = E + R + \Delta S \qquad (1)$$

152 where  $\Delta S$  is the change in water storage including that in soil moisture, snow, and the canopy  
153 water storage. All terms are expressed in depth per unit time units (e.g. mm/year). When a  
154 system is in equilibrium, at annual or longer timescales  $\Delta S = 0$  and  $P = E + R$ .  $\Delta S$ , however,  
155 may not be zero even over long timescales when a system is not in equilibrium e.g., when snow  
156 is accumulating or is melting consistently. We evaluated the P, E, and R components of equation  
157 (1) simulated by CanRCM4 for each of the six river basins, considered in this analysis, and routed  
158 R to obtain streamflow at the river mouths.

159

### 160 **2.1 The Canadian Regional Climate Model (CanRCM4)**

161 CanRCM4 uses the fourth-generation Canadian atmospheric physics (CanAM4) package  
162 (von Salzen et al., 2013), which is the product of a multi-decadal program of climate model  
163 development at the Canadian Centre for Climate Modelling and Analysis (CCCma), a section



164 within Environment and Climate Change Canada. The CanAM4 atmospheric physics package is  
165 also used in CanESM2 (Arora et al., 2011) which contributed results to CMIP5. The difference  
166 between CanRCM4 and CanESM2, other than the former being a regional climate model and the  
167 latter being a comprehensive global ESM, is that CanRCM4 employs the limited-area  
168 configuration of the Global Environmental Multiscale (GEM) model (Côté et al., 1998), which uses  
169 a semi-Lagrangian dynamical core for advection in the atmosphere and is developed by  
170 Environment and Climate Change Canada's Recherche en Prévision Numérique (RPN) where it is  
171 used both for global and regional numerical weather prediction. CanESM2 on the other hand  
172 uses a spectral dynamical core for advection in the atmosphere. CanRCM4 is driven at its  
173 boundaries with data from its parent model (CanESM2). An overview and technical details of the  
174 coordinated global and regional climate modelling effort used to develop the CanESM2-CanRCM4  
175 system are described in detail by Scinocca et al. (2016). Results from the model's North American  
176  $0.22^\circ$  domain, for a single ensemble member, are primarily used here. In addition, we also used  
177 runoff from CanRCM4  $0.44^\circ$  resolution simulations for the North American domain because of  
178 the availability of a large ensemble (LE) of 50 members (CanRCM4 LE) (ECCC, 2018). The large  
179 ensemble simulations allow the consideration of CanRCM4's internal variability, which is an  
180 intrinsic property of the climate system and models, that is largely irreducible and could account  
181 for a large fraction of the inter-climate model spread (Deser et al., 2020). The results used here  
182 from CanRCM4 form part of its contribution to the coordinated regional climate downscaling  
183 experiment (CORDEX) effort. The North American domain of CanRCM uses a rotated latitude-  
184 longitude projection with the North Pole at longitude  $83^\circ$  E and latitude  $42.5^\circ$  N, as opposed to  
185 the geographic North Pole (longitude  $0^\circ$ E, latitude  $90^\circ$  N).

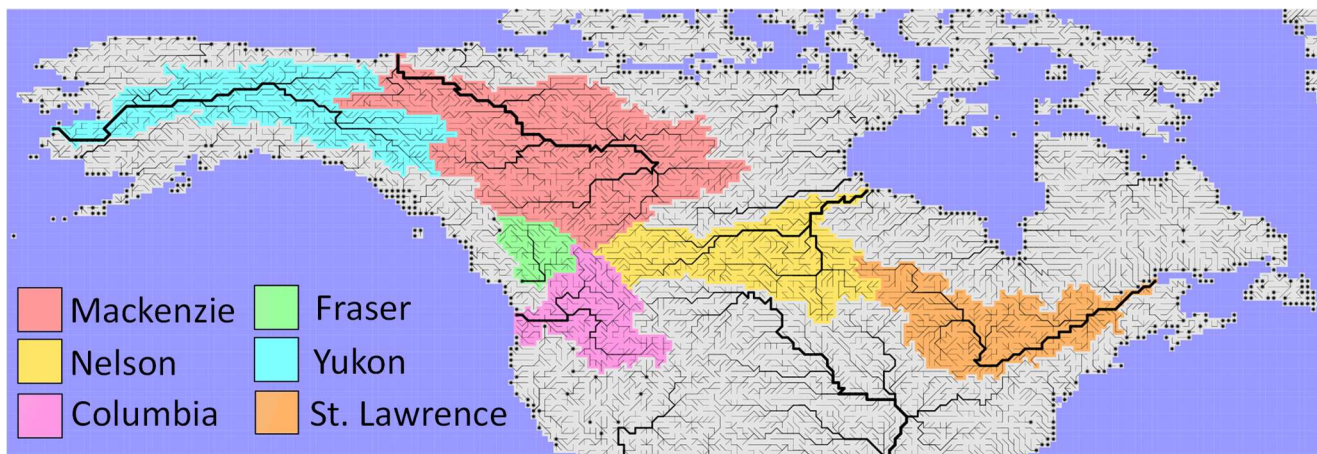
186           The land surface component in CanAM4 is the coupled CLASS-CTEM model. The physical  
187 processes are based on the Canadian Land Surface Scheme (CLASS) (Verseghy, 1991; Verseghy et  
188 al., 1993), and biogeochemical processes (which simulate vegetation as a dynamic component of  
189 the climate system) are based on the Canadian Terrestrial Ecosystem Model (CTEM) (Arora and  
190 Boer, 2003, 2005). The configuration of CLASS-CTEM used in CanESM2 and CanRCM4 uses three  
191 soil layers with thicknesses of 0.10, 0.25, and 3.75 m. Liquid and frozen soil moisture contents,  
192 and soil temperature, are determined prognostically for the three soil layers. The temperature,  
193 albedo, mass, and density of a single-layer snow pack (when environmental conditions permit  
194 snow to exist) are also prognostically modelled. Surface runoff is generated in CLASS when  
195 precipitation intensity exceeds infiltration capacity and when the top soil layer is saturated. The  
196 rainwater and snow melt that infiltrate the soil are available for soil evaporation and  
197 transpiration. Any remaining water percolates down the soil profile and comes out at the bottom  
198 of the soil profile and is termed drainage. Combined surface runoff and drainage constitute total  
199 runoff. Like most land surface components of ESMs, CLASS does not include a groundwater  
200 representation. Surface runoff and drainage from CLASS are used as input into a large-scale river  
201 routing scheme to route runoff and obtain streamflow at the mouth of the rivers considered in  
202 this study as explained in the next section.

## 203 **2.2 Variable velocity routing model**

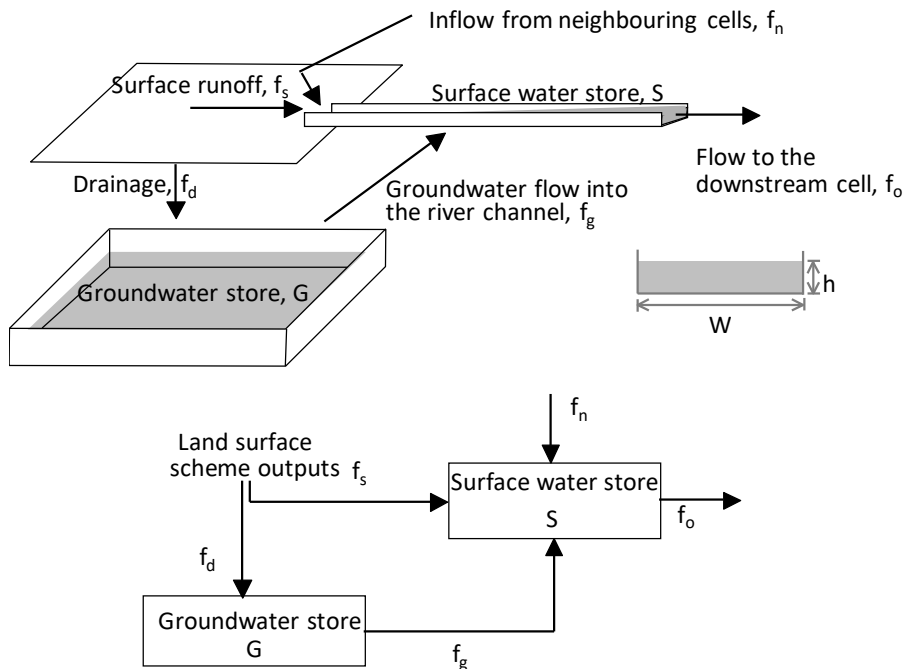
204           The variable velocity river routing scheme of Arora and Boer (1999) that is implemented  
205 in the family of Canadian ESMs (CanESMs) (Arora et al., 2009, 2011; Swart et al., 2019) is used to  
206 route daily runoff from CanRCM4. This routing scheme has been implemented in various versions  
207 of CanESMs at a spatial resolution of  $2.81^\circ$  since the year 2000. For this study, the routing scheme

208 was implemented at a spatial resolution of  $0.5^\circ$ . The reason for using river routing at  $0.5^\circ$   
209 resolution instead of scaling river networks to the  $0.22^\circ$  rotated latitude-longitude projection of  
210 CanRCM4 's North American domain is that scaling river networks is a non-trivial and  
211 cumbersome task that cannot be fully automated (Arora and Harrison, 2007). In contrast,  
212 conservatively regridding runoff from one spatial resolution to another is a straightforward  
213 process. In addition, it has been shown that routing is not very sensitive to the spatial scale at  
214 which it is performed. Specifically, Arora et al. (2001) evaluated the Arora and Boer (1999) routing  
215 scheme together with the WATROUTE routing scheme at  $\sim 350$  km and  $\sim 25$  km spatial resolutions,  
216 respectively, for the Mackenzie River basin. The two routing schemes were driven with the same  
217 runoff. Arora et al. (2001) conclude that for the purpose of realistically modelling streamflow at  
218 the mouth of the rivers in climate models, flow routing at large spatial scales gives similar results  
219 to routing at finer spatial scales. In our study, the difference between the spatial resolution of  
220 runoff ( $0.22^\circ$  and  $0.44^\circ$ ) from the CanRCM4 model and routing ( $0.5^\circ$ ) is much smaller than the  
221 Arora et al. (2001) study. As a result, we do not expect that routing at a slightly different spatial  
222 resolution than runoff will lead to significant differences in the simulated streamflow. The routing  
223 scheme needs river flow directions and these are obtained from the Total Integrating Runoff  
224 Pathways (TRIP) data set (<http://hydro.iis.u-tokyo.ac.jp/~taikan/TRIPDATA/TRIPDATA.html>, last  
225 accessed July 2023) of Oki and Sud (1998). The TRIP data are available at the regular latitude-  
226 longitude grid with the geographic North Pole at its usual location ( $0^\circ$  E,  $90^\circ$  N). Figure 1 shows  
227 the river networks at  $0.5^\circ$  resolution based on TRIP data which also identifies the six river basins  
228 investigated in this study. The Fraser River (identified by the light green colour) appears to have  
229 a river mouth over land. This is because the Fraser River drains into the narrow Strait of Georgia

230 which is not resolved at the  $0.5^\circ$  resolution of the TRIP data set. In addition, the TRIP data set  
 231 does not resolve any inland lakes and provides river flow directions over grid cells that are lakes.  
 232 This is in fact helpful because it avoids discontinuities in the river network.



233  
 234 **Figure 1:** River flow networks at  $0.5^\circ$  resolution used in this study. The major river basins for  
 235 which streamflow and runoff are analyzed in this study are also identified.



236  
 237  
 238 **Figure 2:** Schematic of the Arora and Boer (1999) river routing scheme used in this study to route  
 239 runoff simulated by CanRCM4.

240

241 Figure 2 shows the schematic of the routing scheme which uses surface runoff and  
242 drainage outputs from the land surface scheme. The variable velocity routing scheme used here  
243 is described briefly below and more details can be found in Arora and Boer (1999). The water  
244 balance within a grid cell for its surface  $S$  ( $\text{m}^3$ ) and groundwater  $G$  ( $\text{m}^3$ ) stores is given by

245 
$$\frac{dS}{dt} = f_s + f_n + f_g - f_o \quad (2)$$

246 
$$\frac{dG}{dt} = f_d - f_g \quad (3)$$

247 where,  $f_s$  and  $f_d$  are the surface runoff and drainage (or baseflow) estimates given by the land  
248 surface scheme,  $f_n$  and  $f_o$  are the surface water inflow from the adjacent upstream neighbouring  
249 grid cell(s) and outflow to the downstream grid cell respectively, and  $f_g$  is the groundwater  
250 outflow from the groundwater reservoir to the surface water reservoir within a grid cell as shown  
251 in Figure 2. The fluxes are represented in  $\text{m}^3/\text{s}$ .

252 A river channel is assumed to be rectangular and the width ( $W$ ) of the river at every point  
253 along the river network is specified a priori. This river width in meters is calculated based on its  
254 geomorphological relationship with mean annual discharge. The surface runoff contributes  
255 directly to the surface water store which is essentially the amount of water in the rectangular  
256 river channel between two grid cells. The flow velocity ( $V$ ,  $\text{m}/\text{s}$ ) is calculated using the Mannings  
257 formula (Manning, 1891).

258 
$$V = \frac{1}{r} R^{2/3} S^{1/2} = \frac{1}{r} \left( \frac{A}{P} \right)^{2/3} n^{1/2} = \frac{1}{r} \left( \frac{Wh}{W+2h} \right)^{2/3} n^{1/2} \quad (4)$$

259 where  $r$  is the unitless Mannings roughness coefficient (a default value of 0.04 is used),  $A$  is the  
 260 area of the river channel ( $m^2$ ),  $P$  is the wetted perimeter (m), and  $h$  is the depth of water in the  
 261 channel (m). The slope  $n$  (unitless) of the channel is calculated using elevation difference and the  
 262 river length between two grid cells.

263 The river channel storage  $S$  is assumed to be a linear function of outflow discharge, so

$$264 \quad S = \tau f_o = \frac{L}{V} AV = LA = LWh \quad (5)$$

265 where  $\tau$  is the travel time between the grid cell under consideration and its downstream  
 266 neighbor given by  $\tau = L/V$ , where  $L$  is the distance between the grid cells (m). The outflow  $f_o$  is  
 267 given by

$$268 \quad f_o = AV = WhV = Wh \frac{1}{r} \left( \frac{Wh}{W+2h} \right)^{2/3} n^{1/2} \quad (6)$$

269 and substituting (5) and (6) into (2) yields

$$270 \quad \frac{dh}{dt} = \frac{1}{LW} \left( I - \frac{W^{5/3} h^{5/3}}{r(W+2h)^{2/3}} n^{1/2} \right) \quad (7)$$

271 where  $I$  ( $m^3/s$ ) is the total inflow into a grid cell ( $I = f_s + f_n + f_g$ ). Equation (7) describes the flow in  
 272 terms of the rate of change of flow depth for a given river section. An explicit forward step finite  
 273 difference approximation for (7) yields

$$274 \quad h(t+1) = h(t) + \frac{\Delta t}{LW} \left( I(t) - \frac{W^{5/3} h(t)^{5/3}}{r(W+2h(t))^{2/3}} n^{1/2} \right) \quad (8)$$

275 Flow velocity and outflow discharge for the river channel at any time step can be obtained using  
 276 equations (4) and (6). For the  $0.5^\circ$  resolution used here, a stable solution of (8) is obtained with

277  $\Delta t$  equal to around 10 minutes. The approach yields dynamically-varying flow depth, velocity,  
278 and discharge through the river channel in response to changing surface and baseflow runoff  
279 inputs from the land surface model.

280 The groundwater component of the routing model assumes that groundwater storage,  $G$ ,  
281 is a linear function of groundwater outflow,  $f_g$ .

$$282 \quad G = \tau_g f_g \quad (9)$$

283 The delay in groundwater storage ( $\tau_g$ ) is based on the dominant soil texture type and is set to 10,  
284 35, and 65 days if the dominant soil type in each grid cell is sand, silt, and clay, respectively,  
285 following Arora and Boer (1999). Substituting  $G$  in equation (3) yields

$$286 \quad \tau_g \frac{df_g}{dt} = f_d - f_g \quad (10)$$

287 and following Arora and Boer (1999) we use the following expression

$$288 \quad f_g(t + 1) = f_g(t) e^{-\Delta t/\tau_g} + \left(1 - e^{-\frac{\Delta t}{\tau_g}}\right) f_d(t) \quad (11)$$

289 to determine discharge from the groundwater reservoir within a grid cell and to step forward in  
290 time, where a time step  $\Delta t$  equal to three hours is used. The simplistic form of equation (11)  
291 allows to use a much larger time step than the time step of 10 minutes required for equation (8).

292 The routing scheme used here does not consider the flow regulation effect of dams and  
293 reservoirs. It, however, does consider the effect of lakes and ice jams in a simple manner. The  
294 global lake data set from Kourzeneva et al. (2012) is used which prescribes the fractional coverage  
295 of sub-grid lakes and the five Laurentian Great Lakes (Lakes Superior, Michigan, Huron, Ontario,

296 and Erie). In particular, the flow at the mouth of the St. Lawrence River is affected significantly  
297 by the Great Lakes. The hydraulic residence time of water in the Great Lakes varies from about 2  
298 years for Lake Erie to about 200 years for Lake Superior (Quinn, 1992). As a result, even in the  
299 absence of anthropogenic flow regulation for the St. Lawrence River, we expect the streamflow  
300 at its mouth to show very little seasonality compared to the usual spring peak of Canadian rivers  
301 dominated by snowmelt. The simple approach used here delays the streamflow flowing into a  
302 grid cell with a lake fraction greater than 60% using an e-folding time scale of 300 days similar to  
303 the treatment of the groundwater reservoir (Figure 2) (Arora and Boer, 1999). For the St.  
304 Lawrence River, the effect of delay caused by the Great Lakes is much larger than that of the  
305 anthropogenic flow regulation.

306         Ice jams and breakups are complex thermal and mechanical events and therefore  
307 challenging to model. They occur on all Canadian rivers with varying degrees and depend on  
308 winter temperatures, the river bathymetry, and the physical and geomorphological conditions of  
309 rivers (Beltaos, 2000; Prowse, 1986). The winter freezing of river water inevitably leads to a slow  
310 down of river flow velocity. When water cannot move downstream, upstream flooding results.  
311 Here, we have used a simple approach that increases Manning's roughness coefficient for the  
312 Mackenzie and the Yukon Rivers (which are the most northerly and therefore affected the most  
313 by ice jams) for the period January to June. The value of Manning's roughness coefficient is  
314 increased linearly from 0.04 to 0.08 from 1 January to 31 January, kept at 0.08 from 1 Feb to 31  
315 May, and then reduced linearly from 0.08 to 0.04 over the period June 1 to 30 June. Chen and  
316 She (2020) report the trend in river ice breakup dates for the Mackenzie and Yukon Rivers to be  
317 around -0.3 and -1.3 days/decade for the 1950-2016 period, where the negative sign indicates



318 that the ice breakup is occurring earlier. Assuming the same trend, the breakup dates would  
319 occur about 2.5 and 11 days earlier towards the end of this century, respectively, for the  
320 Mackenzie and Yukon rivers. This simple approach reduces the river flow velocity during the  
321 months that are most affected by river ice jams. Although this is not a perfect nor a complete  
322 approach this simple treatment allows to improve the streamflow seasonality for the Mackenzie  
323 and Yukon rivers. For the southerly Fraser and Columbia rivers such treatment was not necessary.  
324 Consideration of a higher roughness coefficient for the St. Lawrence River to account for ice jams  
325 does not affect its streamflow's seasonality (or rather the lack of it) which is overwhelmingly  
326 determined by the delay and storage caused by the Great Lakes.

### 327 **2.3 Modelled and observation-based data**

328 The CMIP5 historical simulation covers the period 1850-2005 and the future scenarios  
329 cover the period 2006-2100. We used daily runoff from CanRCM4 from its 0.22° North American  
330 domain for the 20 years 1986-2005 from one ensemble member of the historical simulation and  
331 for the 20 years 2081-2100 from one ensemble member each for the two future scenarios (RCP  
332 4.5 and RCP 8.5, Moss et al. (2010)). The RCP 8.5 is the highest baseline emissions scenario where  
333 future development is based on continuous fossil-fuel development. As a result, CO<sub>2</sub> emissions  
334 and concentrations increase throughout the 21<sup>st</sup> century and CO<sub>2</sub> concentration in the year 2100  
335 is around 1100 ppm. RCP 4.5 is a moderate emissions scenario in which emissions peak around  
336 2040 and then decline: as a result CO<sub>2</sub> somewhat stabilizes to around 550 ppm by the year 2100.  
337 Since the CanRCM4 data are available on a rotated latitude-longitude grid and the river routing  
338 is performed on a regular latitude-longitude grid (following the TRIP data), the runoff data from  
339 CanRCM4 are conservatively regridded to the global 0.5° grid using climate data operators (CDO)

340 (<https://code.mpimet.mpg.de/projects/cdo/embedded/index.html#x1-7170002.12.5>, last  
341 accessed Dec 2023) as mentioned earlier. These runoff data are then used as input into the  
342 routing model. The 20-year runoff data (1986-2005 for the historical simulation, and 2081-2100  
343 for the future scenarios) are concatenated into a 40-year time series for each simulation  
344 (historical, RCP 4.5, and RCP 8.5). These data are then input into the routing model and the last  
345 20 years of simulated streamflow are analyzed. The 20-year spin-up is sufficient to allow the  
346 surface and groundwater stores to fill up and reach equilibrium. The simulated precipitation and  
347 temperature from CanRCM4 are compared against observation-based data from the CRU TS 4.07  
348 product (Harris et al., 2020).

349         The simulated streamflow is compared against observation-based estimates obtained  
350 from the Global Runoff Data Centre (GRDC) for the stations that are closest to the river mouths.  
351 Table 1 lists the drainage areas of all rivers considered in this study as discretized in the TRIP data  
352 set and at the stations closest to the river mouth. For the Columbia River, which is heavily  
353 regulated, we obtain an estimate of the naturalized flow with no regulation and no irrigation  
354 provided by the Bonville Power Administration (BPA) for the station VAN (near Vancouver,  
355 Washington, USA) ([https://www.bpa.gov/energy-and-services/power/historical-streamflow-](https://www.bpa.gov/energy-and-services/power/historical-streamflow-data)  
356 [data;https://www.bpa.gov/-/media/Aep/power/historical-streamflow-reports/historic-](https://www.bpa.gov/-/media/Aep/power/historical-streamflow-reports/historic-streamflow-nrni-flows-1929-2008-corrected-04-2017.csv)  
357 [streamflow-nrni-flows-1929-2008-corrected-04-2017.csv](https://www.bpa.gov/-/media/Aep/power/historical-streamflow-reports/historic-streamflow-nrni-flows-1929-2008-corrected-04-2017.csv), last accessed July 2023). The drainage  
358 area of the Columbia River upstream of the VAN station is 616960 km<sup>2</sup> and does not include  
359 discharge contributions from three tributaries (Willamette, Cowlitz, and Lewis Rivers). Of these  
360 three tributaries, the contribution from Willamette is the largest. We obtained naturalized  
361 streamflow for the Willamette River at the station SVN (drainage area 25,600 km<sup>2</sup>) also from

362 BPA's website ([https://www.bpa.gov/-/media/Aep/power/historical-streamflow-](https://www.bpa.gov/-/media/Aep/power/historical-streamflow-reports/correction-20220801.zip)  
 363 [reports/correction-20220801.zip](https://www.bpa.gov/-/media/Aep/power/historical-streamflow-reports/correction-20220801.zip), from the file SVN6ARF\_daily\_COR.xlsx) and added it to the  
 364 naturalized streamflow at the station VAN. This yields naturalized streamflow for the entire  
 365 Columbia River basin, except the smaller Cowlitz, and Lewis Rivers, and represents a drainage  
 366 area of 642,560 km<sup>2</sup> (see Table 1).

367 The Nelson River is affected by two large lakes, Lake Winnipeg and Lake Manitoba, and it  
 368 is also heavily regulated. It currently has five dams towards the end of its journey as it flows into  
 369 Hudson Bay. There are no upstream gauging stations close to the first upstream dam. In addition,  
 370 water is also diverted from Churchill to the Nelson River. We were unable to obtain naturalized  
 371 flow for the Nelson River from the Manitoba hydroelectricity company. Due to anthropogenic  
 372 flow regulation on the Nelson River, the present-day streamflow shows very little seasonality (as  
 373 shown later). As a result, we do not evaluate the simulated daily or monthly streamflow for the  
 374 Nelson River and focus only on its mean annual value.

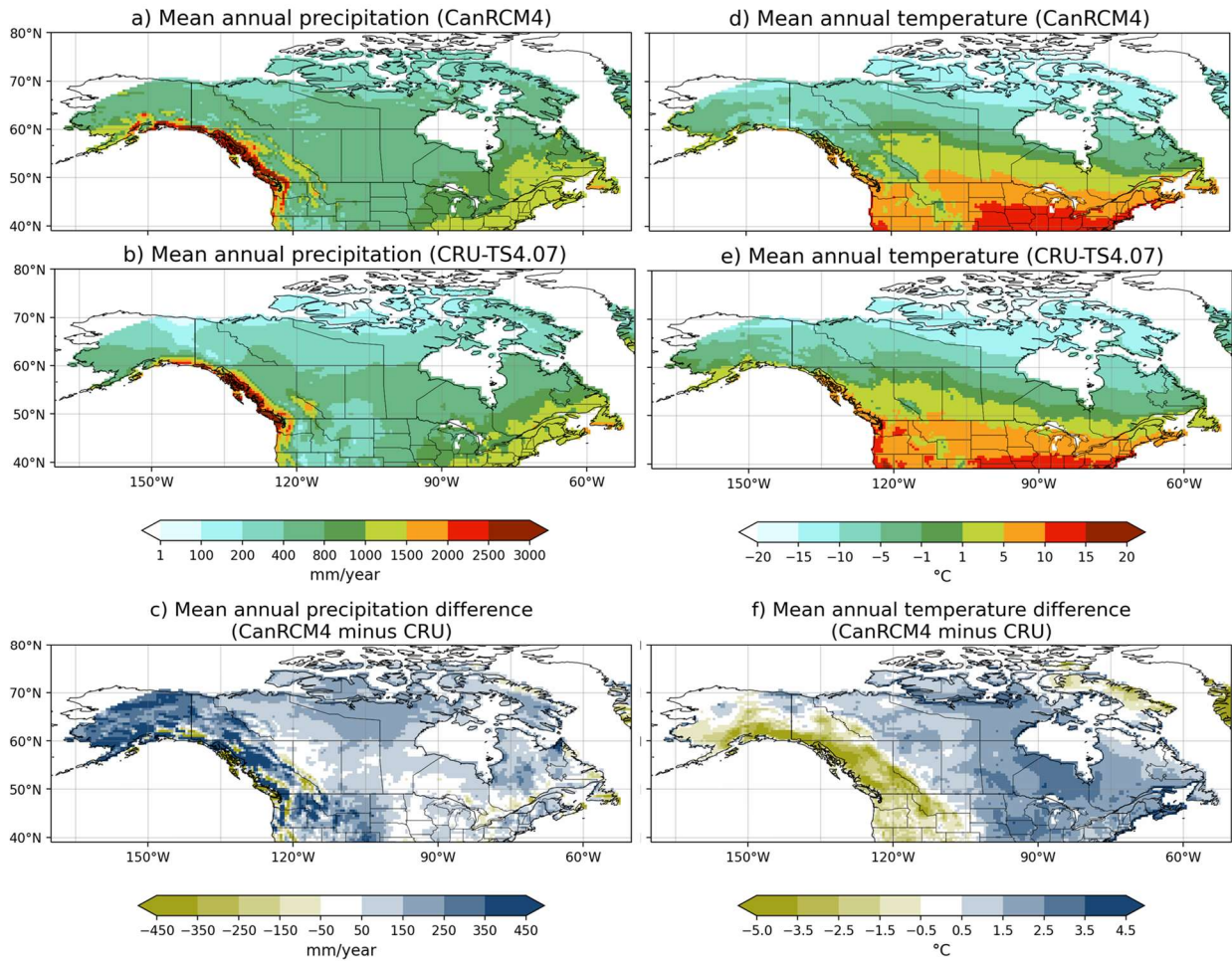
375 **Table 1:** Comparison of river basin areas as represented in the TRIP data and at the gauging  
 376 station closest to the river mouth for the river basins considered in this study as obtained from  
 377 the GRDC.

378

River basin	River basin area (million km <sup>2</sup> )		Gauging station
	in the TRIP data set	at the gauging station closest to the river mouth	
Mackenzie	1.74	1.66	Arctic red river
Yukon	0.85	0.83	Pilot Station
Columbia	0.66	0.64	See section 2.3
Fraser	0.23	0.22	Hope
Nelson	1.07	1.06	Long Spruce generating station
St. Lawrence	1.11	0.77	Cornwall, Ontario

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 382 **Figure 3:** Comparison of CanRCM4 simulated precipitation (left column) and temperature (right  
 383 column) with observation-based estimates from the CRU TS 4.07 data set for the period 1986-  
 384 2005.

385

### 386 3. Results

#### 387 3.1 Present-day precipitation, temperature, and streamflow

388 Figure 3 compares the geographical distribution of mean annual precipitation (left  
 389 column) and temperature (right column) simulated by CanRCM4 to observation-based estimates  
 390 from the CRU TS 4.07 data set (referred to as CRU from here on) for the 1986-2005 period.  
 391 Although the six river basins considered in this study do not cover the entire Canadian region, for

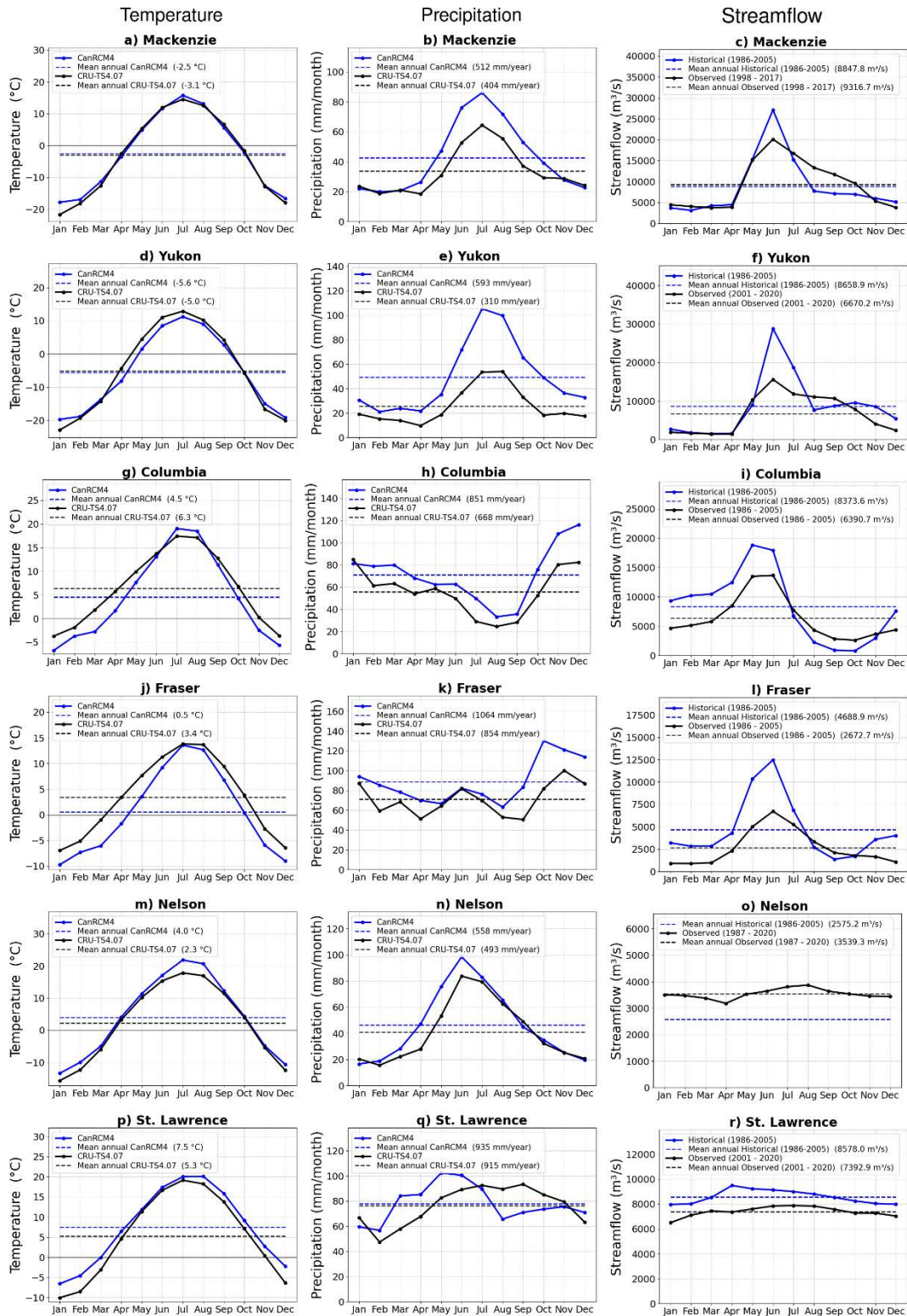
392 completeness the plots are shown for the whole of Canada and south up to 39 °N to include the  
393 southern edge of the Columbia River basin. In Figure 3, while CanRCM4 broadly simulates the  
394 geographical distribution of temperature and precipitation reasonably realistically, there are  
395 differences compared to the CRU data set. CanRCM4 generally simulates higher precipitation  
396 over Canada and more so to the west of the Rockies (Figure 3c) compared to observations. The  
397 model simulates cooler than observed temperatures to the west of the Rockies and higher than  
398 observed temperatures to the east of the Rockies (Figure 3f). This is likely related to the  
399 representation of topography in the model. The overall somewhat higher precipitation in  
400 CanRCM4 over North America is also noted by Alaya et al. (2019) who compared probable  
401 maximum precipitation (PMP) calculated using CanRCM4 data to estimates based on several  
402 reanalyses. Alaya et al. (2019) concluded that among the three reanalyses they considered,  
403 CanRCM4 compared best with the National Centre for Environmental Prediction's (NCEP) Climate  
404 Forecast System Reanalysis.

405 Figure 4 compares the simulated annual cycle of temperature (left column) and  
406 precipitation (middle column) over the six river basins (Figure 1) selected in this study with  
407 observation-based estimates from CRU. The right-hand side column compares simulated  
408 streamflow for the six river basins with observation-based estimates from the GRDC. The basin-  
409 averaged values of temperature and precipitation are calculated by area weighting the values in  
410 the individual grid cells that lie inside a given river basin according to the TRIP data (Figure 1).  
411 The plots also show the mean annual values (dashed lines) on the plot and their magnitude in  
412 the legend. Figure 4 shows that overall CanRCM4 simulated basin-wide averaged temperatures  
413 compare reasonably well with observation-based estimates based on the CRU data for the

414 Mackenzie and the Yukon River basins. For the Columbia and Fraser, the simulated temperatures  
415 are lower for most months, and for the Nelson River basin, the CanRCM4 simulated temperatures  
416 are higher compared to the CRU data. The seasonal cycle of temperature compares well with the  
417 observation-based estimates from CRU data. Compared to temperature, there are larger  
418 differences in simulated CanRCM4 precipitation compared to the CRU data. Although CanRCM4  
419 simulates the seasonality of precipitation reasonably well compared to the CRU data, simulated  
420 precipitation is higher for all river basins, consistent with Figure 3c. The comparison with the CRU  
421 data provides useful insights into simulated quantities. Specifically, despite the difference in the  
422 magnitudes, CanRCM4 provides a reasonable representation of the seasonality of precipitation,  
423 for example, higher winter precipitation in the southern Fraser and Columbia basins, and higher  
424 summer precipitation in the northern Mackenzie and Yukon basins. However, all observation-  
425 based data sets (including CRU) have their limitations. Wong et al. (2017) compared several  
426 gridded observation-based precipitation data sets over Canada and found that they all have  
427 limitations and the data sets compared best with gauge-based precipitation data in summer,  
428 followed by autumn, spring, and winter in order of decreasing quality. Sun et al. (2018) compare  
429 global precipitation from 22 gauge-, satellite-, and reanalysis-based products, including CRU, and  
430 quantify the uncertainty in the different precipitation estimates over timescales ranging from  
431 daily to annual. Shi et al. (2017) evaluated the CRU precipitation over large regions of China and  
432 found that CRU underestimates precipitation in that region compared to rain gauge records.  
433 Furthermore, observation-based precipitation datasets also generally tend to underrepresent  
434 total precipitation in mountainous western Canada (where the Yukon, Mackenzie, Fraser, and  
435 Columbia River basins are located) due to low station density at high elevations (Werner et al.,

436 2019). In the end, the objective of the comparison of the simulated climate with CRU  
437 observations is to evaluate if the model climate is reasonably realistic for the present day. The  
438 assumption behind using direct output from climate models is that despite the biases in the  
439 simulated current climate it is possible to deduce meaningful information about the effect of  
440 climate change using the change in simulated quantities.

441



442  
 443 **Figure 4:** Comparison of the annual cycle of basin-wide averaged CanRCM4 simulated  
 444 temperature (left column) and precipitation (middle column) with observation-based estimates  
 445 from the CRU TS 4.07 data set for the period 1986-2005. The right-hand side column compares



446 simulated streamflow with observations from the GRDC. In the absence of the consideration of  
447 anthropogenic flow regulation for the Nelson River only its simulated mean annual streamflow  
448 value is evaluated.

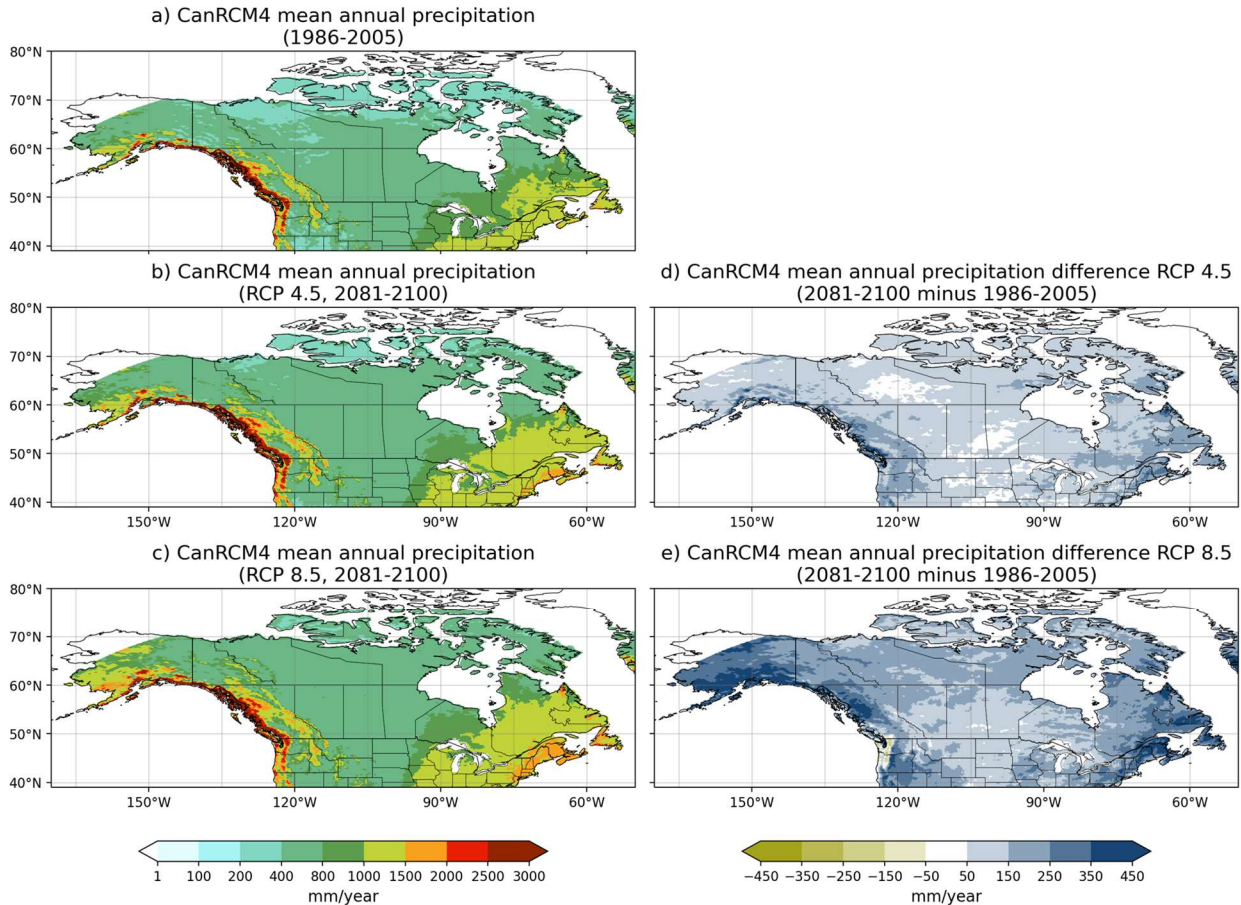
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450           The differences in simulated climate between CanRCM4 and the observation-based  
451 climate in CRU for the present day affect simulated streamflow as expected. The simulated mean  
452 annual streamflow is higher for four out of six river basins considered (Yukon, Columbia, Fraser,  
453 and St. Lawrence) primarily because of the higher simulated precipitation. Simulated  
454 precipitation is also higher for the Mackenzie River basin, but the mean annual simulated  
455 streamflow compares well with its observation-based estimate. Possible reasons for reasonably  
456 realistic annual simulated streamflow despite higher precipitation could be biases in the CRU  
457 data set itself (e.g., underrepresentation of total annual precipitation), or higher simulated  
458 evaporation in CanRCM4 (although simulated summer temperatures compare well with the CRU  
459 data). Finally, the simulated mean annual streamflow for the Nelson River is lower than its  
460 observation-based estimate despite somewhat higher simulated precipitation than the CRU data.  
461 The most likely reason for this is the diversion from the Churchill River into the Nelson River which  
462 started in 1976 to increase the water flow to larger generating stations on the lower Nelson River.  
463 The Manitoba government estimates that an average of 25% more water flows into the lower  
464 Nelson River due to the Churchill River Diversion (CRD)  
465 (<https://www.gov.mb.ca/sd/water/water-power/churchill/index.html>, last accessed Sep. 2023).  
466 The seasonality of streamflow for the Mackenzie, Yukon, and Fraser Rivers is dominated by the  
467 spring snowmelt with the peak occurring in June for both simulated and observed streamflow.  
468 The simulated streamflow for the Columbia and Fraser rivers peaks at the right time but there is  
469 more simulated streamflow during the winter months when precipitation is also higher than

470 observed. For the Mackenzie and Yukon rivers although the mean annual simulated and observed  
471 streamflow are comparable their seasonal distribution is not. The simulated streamflow peak for  
472 these rivers is higher due to the simple treatment of ice jams which is not sufficient to hold the  
473 water in the river channel and then release it slowly as ice jams slowly dissipate in the spring and  
474 summer months, as the observed streamflow indicates. Finally, for the St. Lawrence River, there  
475 is little seasonality in observed streamflow due to the delay caused by the Great Lakes and  
476 anthropogenic flow regulation. The lack of strong seasonality simulated in simulated streamflow  
477 for the St. Lawrence River is caused entirely due to the delay caused by the Great Lakes (section  
478 2.2).

479 Overall the spatial distribution of precipitation and temperature over Canada (Figure 3),  
480 and the seasonality of these two primary climate drivers for the river basins considered in this  
481 study (Figure 4), compare reasonably well with observation-based estimates from the CRU data,  
482 although there are differences in the absolute magnitude of these variables. The resulting  
483 seasonality of streamflow has limitations due to three factors: 1) the biases in the driving climate  
484 from CanRCM4, 2) the biases in the land surface component of CanRCM4 which partitions  
485 precipitation into evaporation and runoff, 3) the lack of calibration of the land surface component  
486 to specific river basins, and 4) the lack of processes in the routing component including the  
487 limitation of not being able to treat ice jams comprehensively. Despite these limitations, the  
488 simulated streamflow captures the broad seasonal patterns with higher values during the spring  
489 snow melt and lower values during the winter months as observations show.

490



491 **Figure 5:** Comparison of CanRCM4 simulated precipitation for the 1986-2005 and for the 2081-  
 492 2100 periods for RCP 4.5 and 8.5 scenarios.

493

### 494 3.2 Changes in future climate and streamflow

495 Figures 5, 6, and 7 show the changes in CanRCM4 simulated precipitation, temperature,  
 496 and runoff for the period 2081-2100, for both RCP 4.5 and 8.5 scenarios, compared to the 1986-  
 497 2005 period from the historical simulation. Over Canada, simulated precipitation and  
 498 temperature increase almost everywhere and in both scenarios. As expected, the magnitude of  
 499 precipitation and temperature change is higher for the RCP 8.5 than the RCP 4.5 scenario.  
 500 Simulated precipitation increases are higher in the coastal western and eastern Canadian regions  
 501 than in central and northern parts of Canada. The central Canadian region sees the lowest

502 increase in precipitation in both scenarios. Simulated temperature increases, as expected, are  
503 higher at higher latitudes due to polar amplification of the temperature change associated with  
504 the snow- and ice-albedo feedbacks. In the RCP 4.5 and 8.5 scenarios, the simulated temperature  
505 changes vary from about 3 °C and 6 °C, respectively, in the south, to about 6 °C and 11 °C, in the  
506 north. The parent climate model (CanESM2) on which CanRCM4 is based has an equilibrium  
507 climate sensitivity of 3.7 °C, somewhat on the higher side, compared to the range of 1.5 °C to 4.5  
508 °C amongst climate models that contributed to CMIP5 (Schlund et al., 2020). As a result, we also  
509 expect the magnitude of simulated changes to be somewhat higher than a model with average  
510 climate sensitivity.

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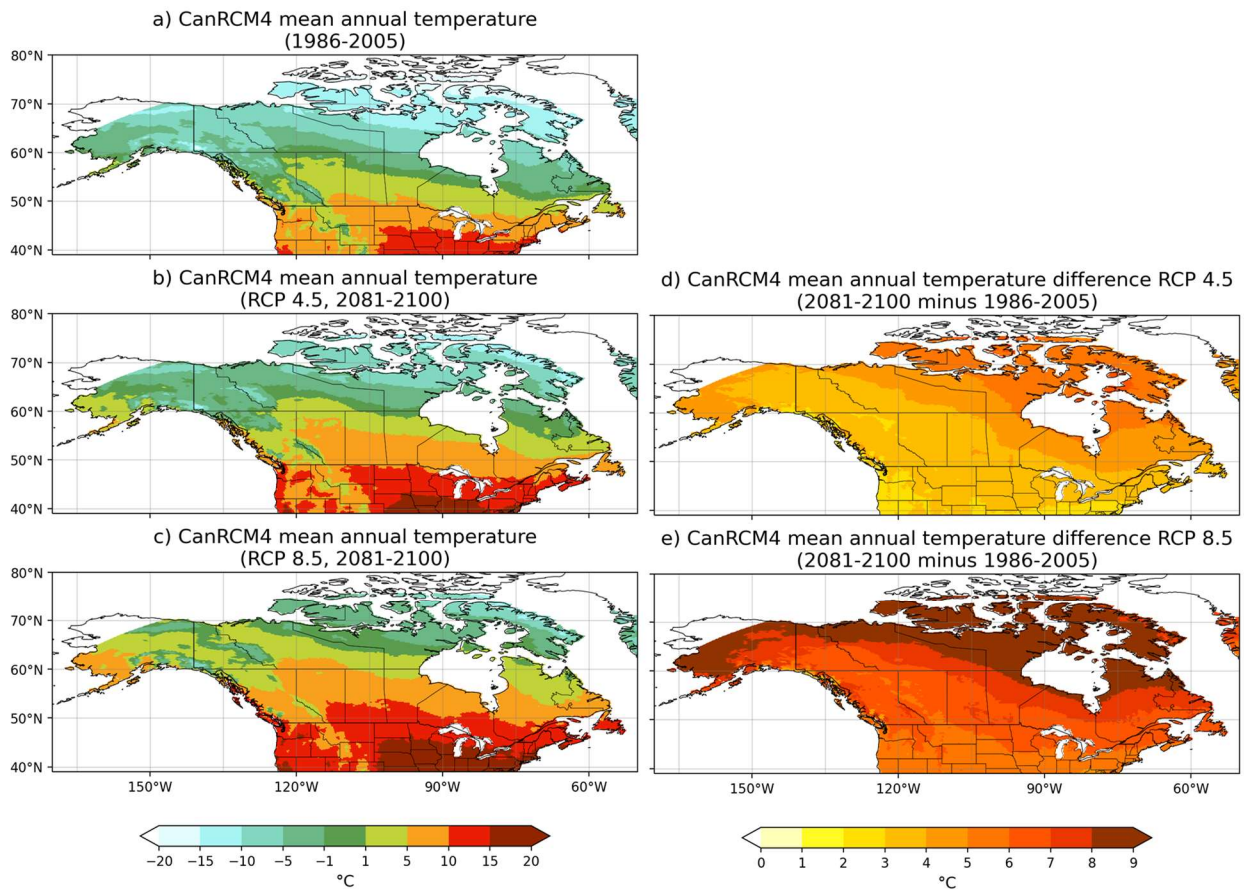
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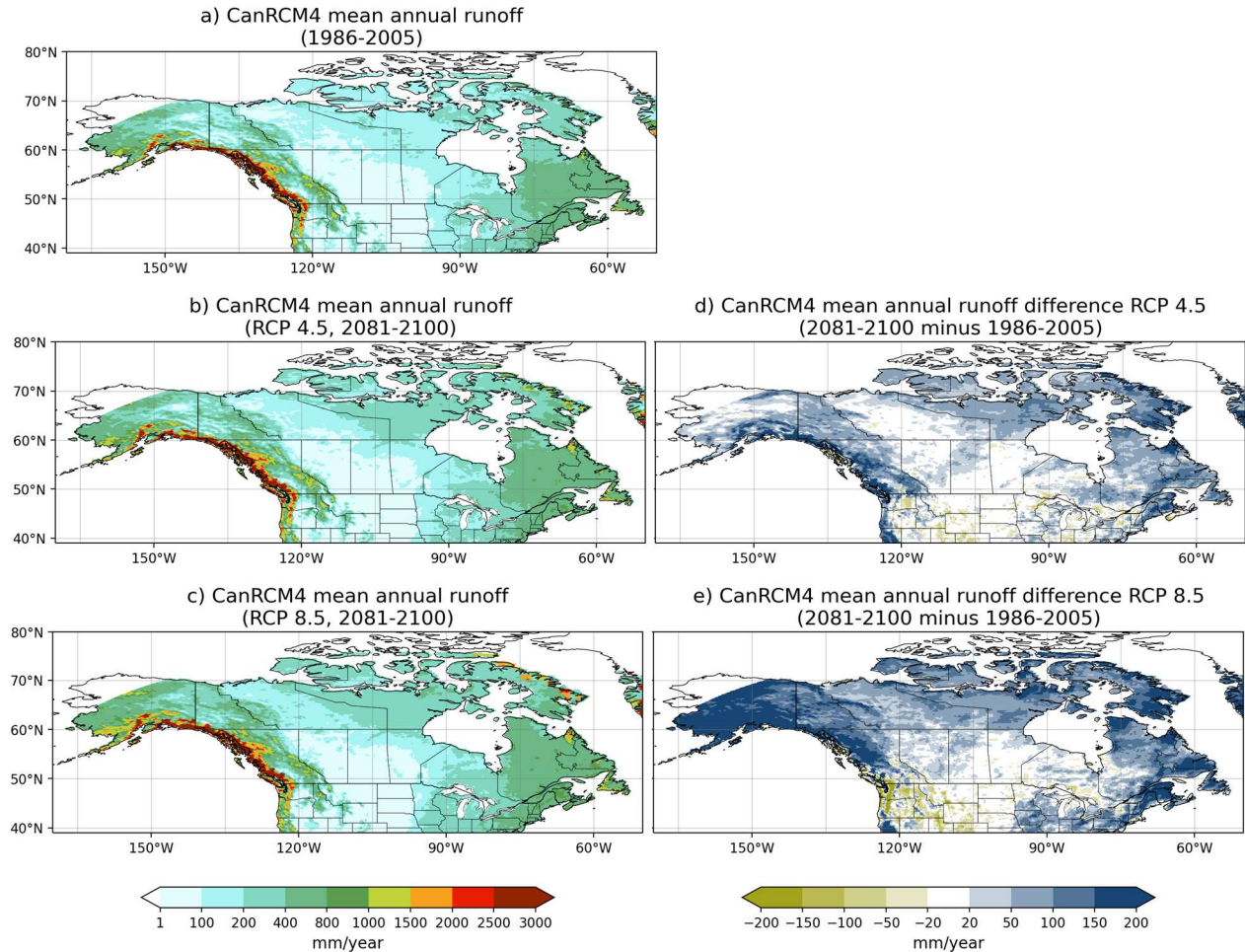
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519 **Figure 6:** Comparison of CanRCM4 simulated temperature for the 1986-2005 period and for the  
520 2081-2100 periods, for RCP 4.5 and 8.5 scenarios.



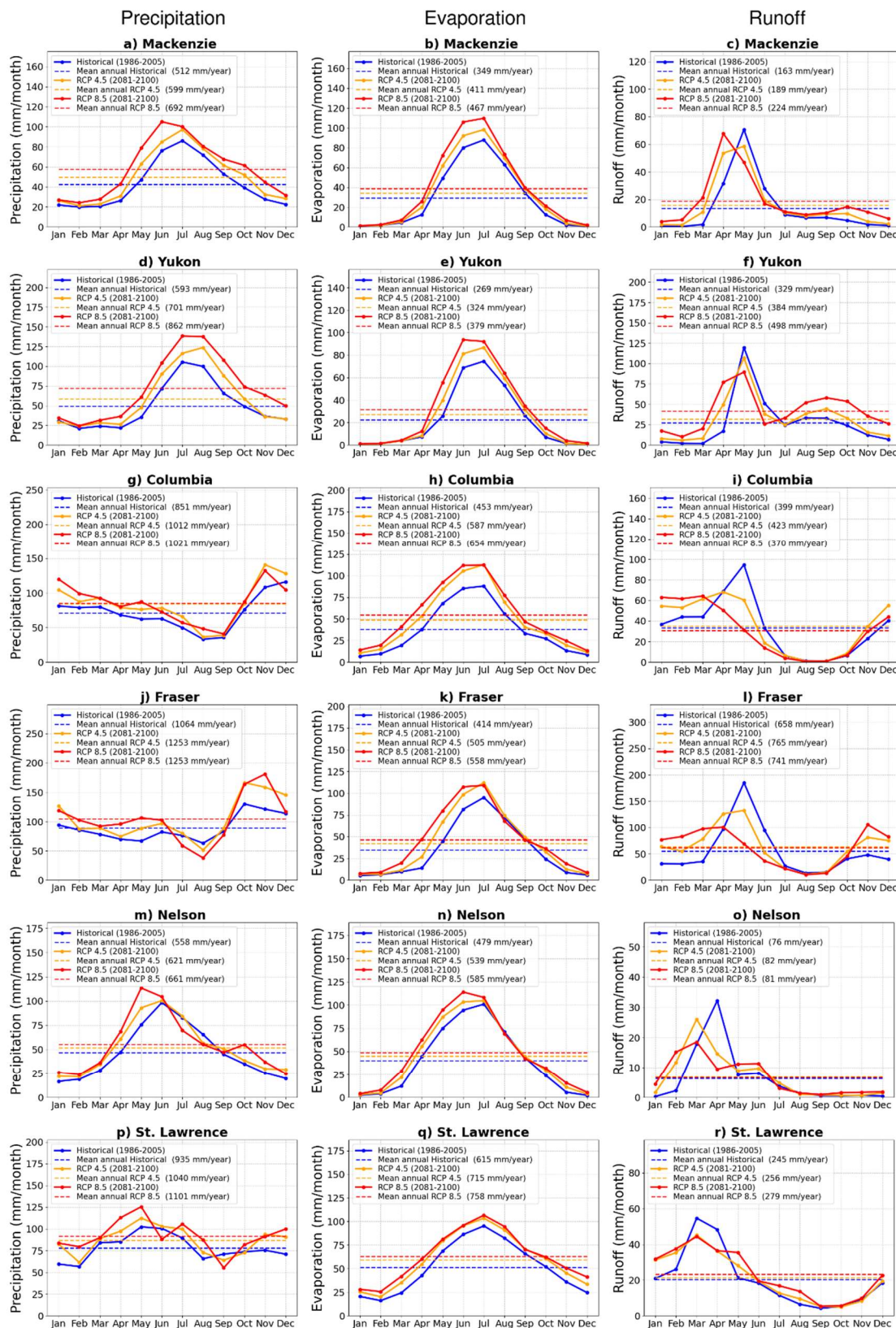
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523 **Figure 7:** Comparison of CanRCM4 simulated runoff for the 1986-2005 period and the 2081-  
 524 2100 periods, for RCP 4.5 and 8.5 scenarios.

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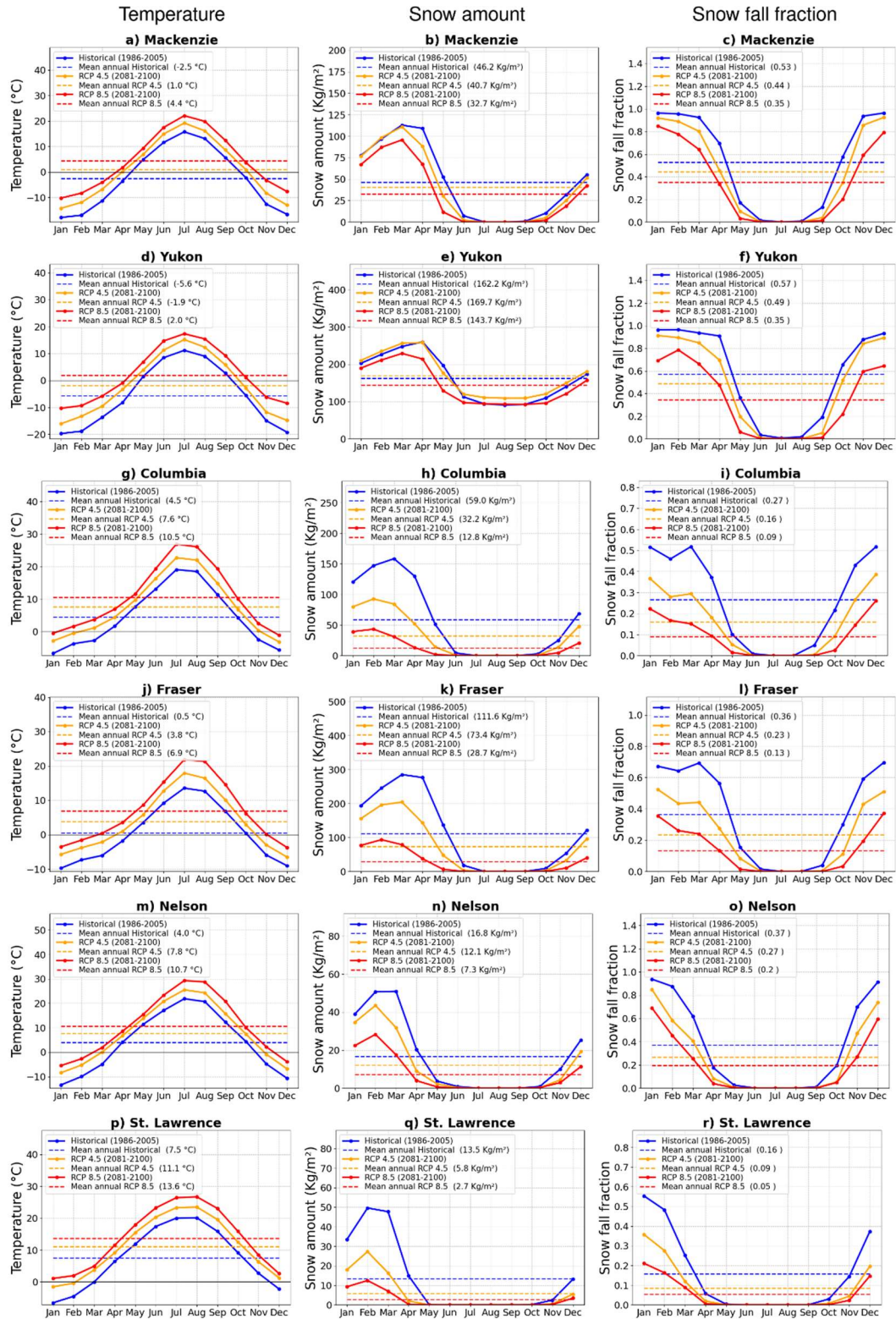
526 In Figure 7 runoff increases generally everywhere in Canada for the RCP 4.5 and RCP 8.5 scenarios  
 527 with larger changes on the west and east coasts, and in northern Canada, following a similar  
 528 pattern of changes in precipitation. Runoff reduces in parts of the southern Columbia River basin  
 529 in the United States in the RCP 4.5 scenario, and these decreases become more pronounced and  
 530 widespread over the north-western Pacific region in the RCP 8.5 scenario including the Fraser  
 531 River basin in Canada.

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562 **Figure 8:** Comparison of the annual cycle of basin-wide averaged CanRCM4 simulated water  
563 budget components for each river basin for the historical (1986-2005) period and the two future  
564 scenarios RCP 4.5 and 8.5 (2081-2100): precipitation (left column), evaporation (middle column),  
565 and runoff (right column).

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**Figure 9:** Comparison of the annual cycle of basin-wide averaged CanRCM4 simulated temperature (left column), snow water equivalent amount (middle column), and snowfall fraction (right column) for the historical (1986-2005) period and the two future scenarios RCP 4.5 and 8.5 (2081-2100).



609 Figure 8 shows the annual cycle of the simulated water budget components  
 610 (precipitation, evaporation, and runoff) for the six river basins considered in this study for the  
 611 historical (1986-2005) period and the two future scenarios, RCP 4.5 and 8.5 (2081-2100). As in  
 612 Figure 4, the mean annual values are shown as dashed lines, and their magnitude is noted in the  
 613 legend.

614 **Table 2:** Evaporation and runoff ratios for the six river basins simulated by CanRCM4 for the  
 615 historical period (1986-2005) and the two future scenarios (RCP 4.5 and 8.5, 2081-2100). The  
 616 evaporation (runoff) ratio is the ratio of mean annual evaporation (runoff) to precipitation.  
 617

River basin	Evaporation ratio (E/P)			Runoff ratio (R/P)		
	Historical (1986-2005)	RCP 4.5 (2081-2100)	RCP 8.5 (2081-2100)	Historical (1986-2005)	RCP 4.5 (2081-2100)	RCP 8.5 (2081-2100)
Mackenzie	0.682	0.686	0.675	0.318	0.316	0.324
Yukon	0.454	0.462	0.440	0.555	0.548	0.579
Columbia	0.532	0.580	0.641	0.469	0.418	0.362
Fraser	0.389	0.403	0.445	0.618	0.611	0.591
Nelson	0.858	0.868	0.885	0.136	0.132	0.123
St. Lawrence	0.664	0.686	0.684	0.314	0.294	0.302

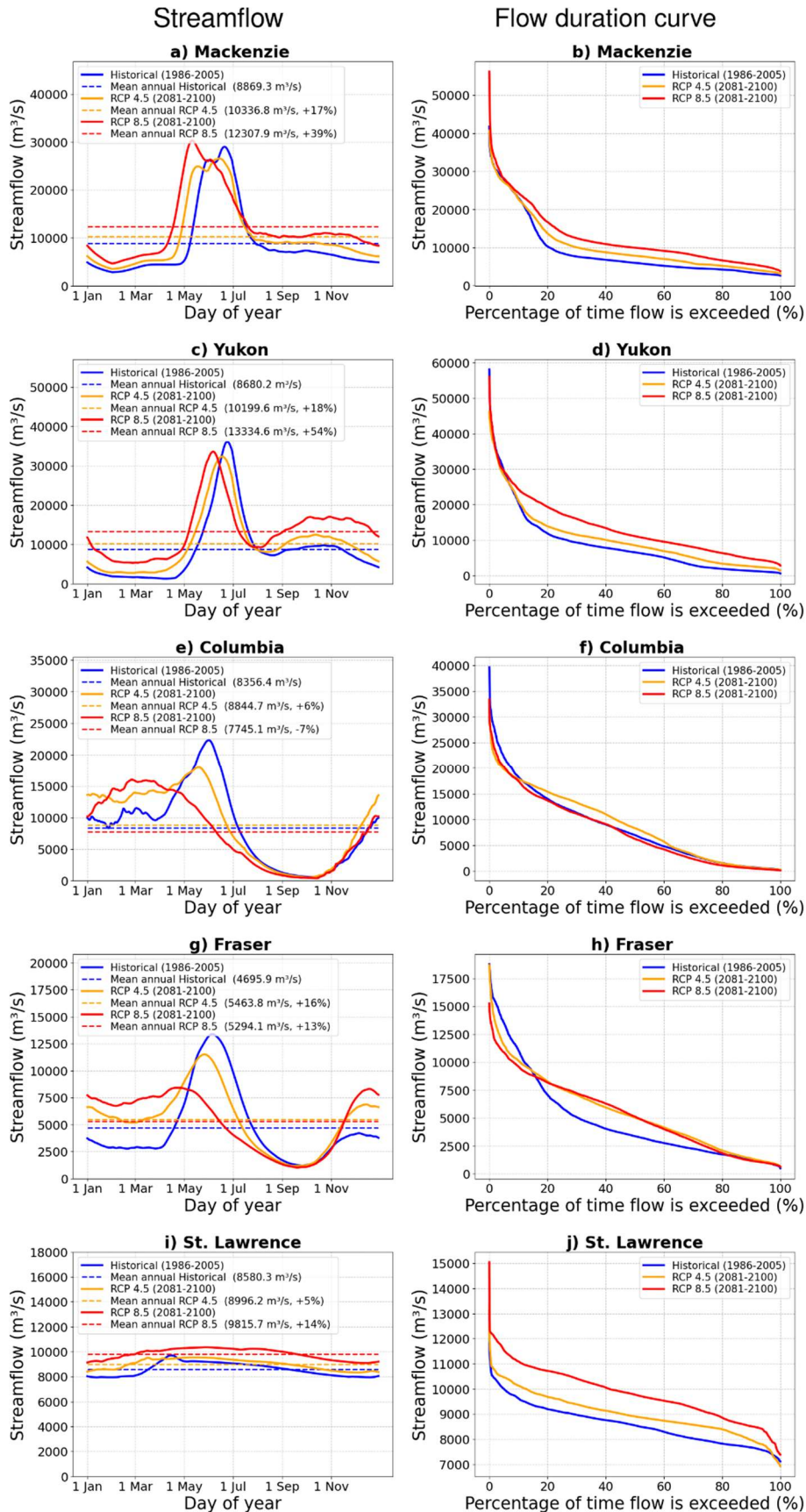
618  
 619 The evaporation (E/P) and runoff (R/P) ratios for the six river basins for the historical period and  
 620 the two future scenarios are shown in Table 2 and allow to see how the partitioning of  
 621 precipitation into evaporation and runoff changes with climate. For the mean annual values of P,  
 622 E, and R reported in Figure 8, P is balanced to within 1% by E+R for all river basins (except St.  
 623 Lawrence) and all scenarios, except for the Yukon (for RCP 8.5) and Fraser River basins (for RCP  
 624 4.5 and 8.5) for which (E+R) is higher than P indicating that  $\Delta S$  is not zero (see equation 1). As a  
 625 result, (E/P) and (R/P) also add to one for all river basins except for the Yukon (RCP  
 626 8.5,  $(E + R)/P = 1.02$ ) and the Fraser River (RCP 4.5,  $(E + R)/P = 1.014$ , and RCP 8.5,  
 627  $(E + R)/P = 1.036$ ) basins. For the St. Lawrence River basin, the imbalance is around 2%  
 628 because of the presence of the Great Lakes which had to be excluded from the river basin mask.

629 Since basin-wide averaged calculations are done at 0.5° latitude-longitude resolution, and the  
630 actual domain of CanRCM4 is on a rotated latitude-longitude projection this led to slightly more  
631 rounding errors for the St. Lawrence than other river basins.

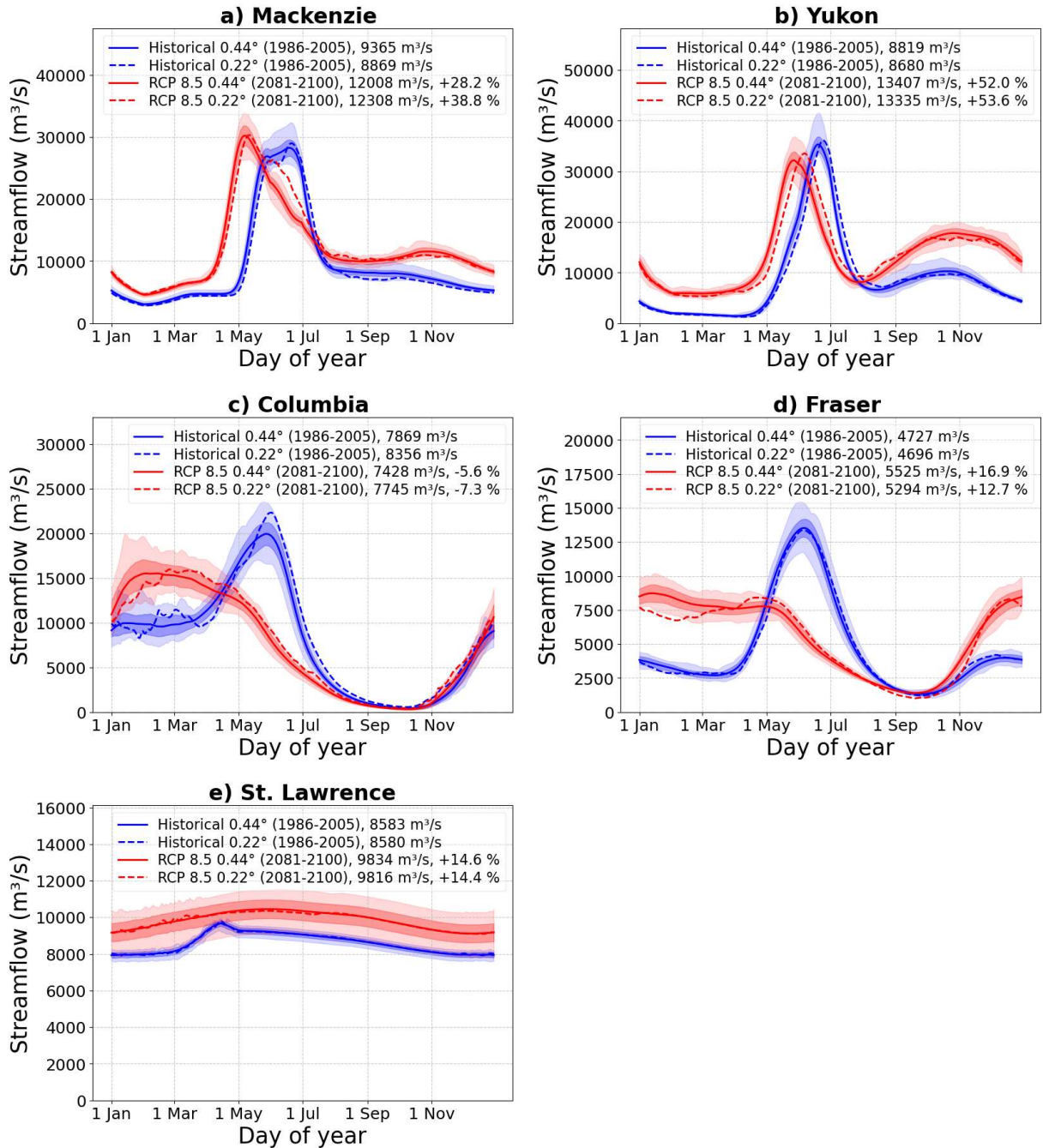
632 For all river basins considered, precipitation increases for both future scenarios with the  
633 increase being larger for the RCP 8.5 scenario consistent with Figures 5d and 5e. The response of  
634 evaporation to changes in climate is expected. The increase in precipitation and temperature  
635 yields an increase in evaporation for future scenarios for all river basins. Simulated runoff does  
636 not increase as much as precipitation since evaporation also increases. The runoff ratio, in Table  
637 2, increases for the northerly Mackenzie and the Yukon River basins while it decreases for the  
638 southerly Nelson, St. Lawrence, and especially for the Fraser and Columbia River basins which  
639 are characterized by milder climate owing to their location in the Pacific north-western region.  
640 This is because the increase in precipitation is more than enough to compensate for the increase  
641 in evaporation (associated with a warmer climate) for the northern river basins but not for the  
642 southern ones (as seen earlier in Figure 7 where runoff begins to decrease in parts of the  
643 Columbia and Fraser River basins). The absolute runoff amount in Figure 8 increases for the  
644 Mackenzie and Yukon River basins, in the RCP 4.5 and 8.5 scenarios compared to the historical  
645 simulation, but doesn't change much for the Columbia, Fraser, Nelson, and St. Lawrence River  
646 basins. However, the seasonality of runoff changes for all river basins, and the peak in simulated  
647 runoff either occurs earlier in the year, occurs with reduced magnitude, or both. Canadian rivers  
648 are dominated by spring snowmelt and this runoff behaviour is associated with snow melt  
649 occurring earlier in the year in the RCP 4.5 scenario than in the historical simulation and occurring  
650 even earlier in the RCP 8.5. This is seen in Figure 9 which shows the simulated annual cycle of

651 temperature changes, snow amount, and snowfall as a fraction of total precipitation for the  
652 historical period and the two RCP scenarios for the six river basins. In Figure 9 the mean annual  
653 temperature increases from the historical period to the RCP 4.5 scenario, and from the RCP 4.5  
654 to RCP 8.5 scenario, are between 3 and 3.5 °C for the six river basins considered here. The middle  
655 column of Figure 9 shows that in addition to earlier snowmelt the amount of snow in the winter  
656 months decreases for all river basins with climate warming. The only exception to this is the  
657 Yukon River basin in which the mean annual snow amount increases marginally in the RCP 4.5  
658 scenario (Figure 9e). As expected, the fraction of precipitation falling as snow also decreases with  
659 climate warming for all river basins (right column, Figure 9).

660



**Figure 10:** Comparison of the simulated daily streamflow (left column) and flow duration curves (right column) for the historical (1986-2005) period and the two future scenarios RCP 4.5 and 8.5 (2081-2100) for the river basins considered. The Nelson River is excluded for which we only evaluated annual streamflow values that are mentioned in the text.



687 **Figure 11:** Comparison of the simulated daily streamflow for the historical (1986-2005) period  
 688 and the RCP 8.5 scenario (2081-2100) for the river basins considered in this study from the  
 689 0.22° and 0.44° simulations. The results from the 0.22° simulations (shown earlier in Figure 10)  
 690 are shown as dashed lines. The uncertainty range for the 0.44° simulations is based on results  
 691 from CanRCM4's 50-member large ensemble. The solid lines indicate the mean across 50  
 692 members the light shading indicates the full range, and the dark shading indicates the mean  $\pm$   
 693 one standard deviation range, for the 0.44° simulations. The Nelson River is excluded for which

694 only annual streamflow values are analyzed.

695

696 Figure 10 compares simulated daily streamflow and flow duration curves averaged over  
697 the historical (1986-2005) period with those averaged over the two future scenarios RCP 4.5 and  
698 8.5 (2081-2100) for the river basins considered here excluding the Nelson River. The flow  
699 duration curves are calculated using daily streamflow values. The legends in Figure 10 for the  
700 streamflow figures in the left column show mean annual values but also the change from the  
701 simulated historical values for the RCP 4.5 and 8.5 scenarios. The mean annual streamflow  
702 increases for all rivers for both the RCP 4.5 and 8.5 scenarios, except for the Columbia River for  
703 the RCP 8.5 scenario (-7%). The increase in simulated annual streamflow is largest for the  
704 Mackenzie (+16%, +39%) and Yukon Rivers (+17%, +53%) for the RCP 4.5 and 8.5 scenarios, due  
705 to higher precipitation increase in these two basins (Figure 8). The increase in annual streamflow  
706 for other rivers is smaller and between 6% and 14%. Daily streamflow and flow duration curves  
707 are not shown for the Nelson River because we do not consider anthropogenic flow regulation,  
708 as mentioned earlier. The simulated mean annual streamflow for the Nelson River increases from  
709 2556.6 m<sup>3</sup>/s (for the 1986-2005 period) to 2774.8 and 2723.8 m<sup>3</sup>/s for the RCP 4.5 (+9%) and 8.5  
710 (+7%) scenarios, respectively (for the period 2081-2100).

711 The changes in streamflow seasonality are larger for the southerly Columbia and Fraser  
712 Rivers than for the northerly Mackenzie and Yukon Rivers. The peak daily streamflow for the  
713 Yukon River still occurs in June given it's the coldest river basin (Figure 4d) and the streamflow  
714 seasonality is still dominated by the spring snowmelt. The simulated daily peak streamflow for  
715 the Yukon River occurs on 24 June for the historical period (1986-2005), and 18 June and 6 June,

716 respectively, for RCP 4.5 and 8.5 scenarios for the period 2081-2100. Streamflow for the Yukon  
717 River also begins to increase earlier due to earlier snowmelt (Figure 9e). While the spring peak  
718 streamflow reduces in both RCP 4.5 and 8.5 scenarios during June and part of July, streamflow  
719 increases for most other months for the Yukon River. The Mackenzie River shows similar  
720 behaviour to the Yukon River in terms of earlier shifts of spring streamflow peaks with climate  
721 warming but the spring peak is higher for the RCP 8.5 scenario. The mean simulated daily peak  
722 streamflow for the Mackenzie River occurs on 21 June for the historical period (1986-2005), and  
723 14 June and 11 May, respectively, for RCP 4.5 and 8.5 scenarios for the period 2081-2100. Similar  
724 to Yukon, although the streamflow is lower for the Mackenzie River during June and part of July,  
725 it increases for most other months. The corresponding changes in streamflow are also seen in  
726 the flow duration curves. For these two rivers the frequency of the occurrence of flows that occur  
727 greater than about 5% of the time in the historical simulation increases in the future. The  
728 Columbia and the Fraser Rivers experience much larger changes in their seasonality as their  
729 primarily snow-dominated flow regimes change to more hybrid flow regimes. The snowmelt-  
730 driven streamflow peak in spring is reduced considerably for future scenarios since a lower  
731 fraction of fall, winter, and spring precipitation falls as snow. As a result, streamflow increases  
732 from October to April since precipitation falling as rain, as opposed to snow, yields runoff that  
733 runs straight into the rivers. Additionally, the large reduction in snowpack volume together with  
734 earlier melt (Figure 9k and 9h) affects the seasonality of the Fraser and Columbia Rivers  
735 streamflow and causes pronounced shifts in peak flows. The mean simulated daily peak  
736 streamflow for the Columbia River occurs on 1 June for the historical period (1986-2005), and 19  
737 May and 25 February, respectively, for RCP 4.5 and 8.5 scenario for the period 2081-2100. For

738 the Fraser River, the mean simulated peak streamflow occurs on 5 June for the historical period  
739 (1986-2005), and 26 May and 21 April, respectively, for RCP 4.5 and 8.5 scenario for the period  
740 2081-2100. The pronounced changes in the Fraser River basin peak flow are apparent in its flow  
741 duration curve (Figure 10h) which shows a decrease (increase) in the frequency of streamflow  
742 events which occurred less (more) than about 16% of the time and result in a more equitable  
743 streamflow regime with a pronounced reduction in its seasonality. The simulated streamflow for  
744 the St. Lawrence River shows very little seasonality and since annual streamflow increases for  
745 both scenarios, the flow duration curve simply moves up (Figure 10j).

### 746 **3.3 Uncertainty in simulated changes in future streamflow**

747 Using the large ensemble simulations that are available for the historical period and the  
748 RCP 8.5 scenario at 0.44 ° resolution we quantified the uncertainty in the simulated streamflow  
749 associated with the internal variability of the CanRCM4 model. Similar to the 0.22° resolution,  
750 we regridded the 0.44° runoff at CanRCM4's rotated latitude longitude projection to 0.5° regular  
751 latitude longitude projection for use as input into the river routing scheme. This is illustrated in  
752 Figure 11 which shows the simulated daily streamflow for all the rivers considered here except  
753 the Nelson River. In Figure 11, the solid lines show the average across the 50 members of the  
754 large ensemble, light shading shows the full range of the results, and dark shading shows the  
755 mean  $\pm$  one standard deviation range (this implies the 16%-84%, i.e. 68%, range when assuming  
756 normally distributed monthly streamflow values). In addition, streamflow from the 0.22°  
757 simulations (from Figure 10) is shown as dashed lines to allow direct comparison of results from  
758 the 0.22° and 0.44° simulations.



759           The changes in simulated streamflow are consistent between the 0.22° and 0.44°  
760 simulations. The results from the 0.44° simulations are also notably smoother compared to the  
761 0.22° simulations since the 0.44° results are also averaged over the 50 ensemble members in  
762 addition to the 20 years. For the most part, the results from the 0.22° simulations lie within the  
763 full range of results from the 0.44° simulations.  
764 This is expected since the driving climate at the boundaries of CanRCM4 based on CanESM2 is  
765 the same in both resolutions. The magnitude of change from the historical to the RCP 8.5 scenario  
766 (see legend for individual rivers) is, however, somewhat different. This is also expected because  
767 the coarser resolution 0.44° simulations are less representative of the basin topography than  
768 the 0.22°simulations. The day of peak streamflow occurs a few days earlier in 0.22° simulations  
769 than in the 0.44° simulations for the Mackenzie and Yukon Rivers. Overall, the large ensemble  
770 from the 0.44° simulations helps to provide context for results from the 0.22° simulations.

771           Overall, despite the differences in the magnitude of changes, the direction and variability  
772 of change obtained from this study is generally consistent with the previous studies using basin-  
773 scale hydrologic models, driven by statistically downscaled and bias-corrected climate model  
774 data, for instance for the Fraser River (Islam et al., 2019; Shrestha et al., 2012), the Columbia  
775 River (Schnorbus et al., 2014) and the Yukon River (Hay and McCabe, 2010). The results presented  
776 here are also comparable to the projections from global and regional scale hydrologic models,  
777 e.g. for the Mackenzie River basin (Krysanova et al., 2017, 2020).

#### 778 **4. Summary and conclusions**

779           This study offers a consistent analysis of results across six river basins in Canada based on  
780 results from the CanRCM4 model. Despite the biases in simulated present-day CanRCM4 climate,  
781 and some differences in the results based on 0.22° and 0.44° simulations, the results provide  
782 useful information about changes in simulated streamflow that is consistent with expectations  
783 of process behaviour in a warmer climate, and with published studies.

784           Neither future precipitation nor temperature changes are uniform across Canada.  
785 Simulated precipitation increases are higher closer to the west and east coasts, and simulated  
786 temperature changes are higher towards the Arctic. Similar to precipitation, runoff changes are  
787 also higher closer to the west and east coasts. The changes in simulated streamflow indicate how  
788 the present-day climate state of river basins plays a role in their response to climate change. The  
789 results yield two broadly distinct responses of monthly streamflow changes to climate warming,  
790 up until the end of this century, for the northerly Mackenzie and Yukon rivers and the southerly  
791 Fraser and Columbia rivers. Despite higher future projected temperature changes in Canada's  
792 north, peak streamflow for the Mackenzie and Yukon rivers is still dominated by the spring  
793 snowmelt. This is because the present-day colder states of these river basins imply that even  
794 after around 6-7 °C warming, the basin-wide average temperatures are cold enough to not  
795 sufficiently change their snowmelt-dominated streamflow regimes. Changes, however, do occur  
796 in streamflow seasonality for these two rivers. Mean peak daily streamflow occurs earlier by  
797 about 6-7 days for the Mackenzie and Yukon Rivers in the RCP 4.5 scenario and about 28 days for  
798 the Mackenzie River and 12 days for the Yukon River for the RCP 8.5 scenario (Figure 10). The  
799 earlier start of the snowmelt is the primary factor for the changes in peak streamflow and its time  
800 of occurrence, while the streamflow increases during the rest of the year (except for June and

801 part of July) are driven by an increase in precipitation. Additionally, a higher fraction of winter  
802 precipitation occurring as rainfall drives the winter streamflow increases. In contrast, the  
803 streamflow seasonality for the southerly Fraser and Columbia rivers is significantly more affected  
804 by warmer temperatures because the mean annual basin-wide temperature for these river basins  
805 is already above 0° C for the historical period. Both these rivers experience pronounced changes  
806 in their streamflow seasonality. The peak daily streamflow for both rivers decreases considerably  
807 and occurs about 45 days earlier for the Fraser River and about 100 days earlier for the Columbia  
808 River in the RCP 8.5 scenario. These results compare reasonably to the 1-2 months earlier peak  
809 in previous studies for the Fraser River (Islam et al., 2019; Shrestha et al., 2012) but are higher  
810 than the two months earlier peak for the Columbia River (Schnorbus et al., 2014) that used results  
811 from multiple climate models. Shrestha et al. (2021a) used CanRCM4 data to evaluate snowpack  
812 response to varying degrees of warming. They found that snowpack reduction using CanRCM4-  
813 LE is higher than the ensemble of results obtained by driving a hydrological model with data from  
814 other climate models (their supplementary information), consistent with CanESM2's higher  
815 climate sensitivity. For the Nelson and the St. Lawrence Rivers which show very little seasonality  
816 the effect of climate change is reflected in the changes in mean annual streamflow.

817         The results presented here also appear to show that the simulated changes in streamflow  
818 are somewhat resolution-dependent. This would be expected especially for topography-  
819 dominated river basins. If a large ensemble of 50 members for the 0.22° resolution was also  
820 available, it would have been easier to draw firm conclusions about the effect of the spatial  
821 resolution on changes in simulated streamflow.

822           There are two primary limitations of the work presented here. First, we use results from  
823 only one climate model. It would have been ideal to use runoff from other regional climate  
824 models to provide an uncertainty range based on the spread across different climate models.  
825 This would have also allowed us to evaluate how the spread across models compares to the  
826 spread across the 50 members of the CanRCM4 large ensemble. Second, the results are based on  
827 direct output from the CanRCM4 climate model and direct climate model output is biased. This  
828 limitation is tied to our methodology. The use of bias-corrected climate data inevitably implies  
829 using a different hydrological model or land surface scheme, than the land surface component of  
830 CanRCM4, and forcing it with bias-corrected climate data to obtain runoff. Finally, there are  
831 uncertainties associated with the routing process itself. As mentioned earlier, the routing scheme  
832 accounts for ice jams in a simplified manner, and anthropogenic flow regulation is not taken into  
833 account. The implicit assumption when using raw climate model output is that, despite the biases  
834 in simulated climate, it is possible to derive useful information about the impact of climate  
835 change on the simulated streamflow and other components of the hydrological budget. The  
836 Canada-wide results presented here have allowed us to differentiate between the hydrological  
837 response of the northerly Mackenzie and Yukon Rivers, and the southerly Fraser and Columbia  
838 Rivers, to climate change in a consistent manner. Furthermore, our results help fill the gaps in  
839 regions across Canada, where no climate model-driven hydrological projections are available.  
840 Within the scope of this study, we have only evaluated streamflow at the mouth of the six major  
841 rivers considered here. The full data set of daily simulated streamflow for the 20-year historical  
842 (1986-2005) and future periods (2081-2100) for the two scenarios, based on runoff from the  
843 0.22° simulations, is made available as detailed in the data availability section.

844 Large ensembles are now becoming more common. The challenge for similar future  
845 studies is to consider the inter-model and intra-model (based on ensemble members of the same  
846 model) spreads in the same framework to derive an uncertainty estimate that takes into account  
847 both types of uncertainties.

848  
849

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860

## 861 **Data availability**

862  
863 The CanRCM4 data from 0.22° simulations used in this study are available from CCCma website  
864 (<https://climate-modelling.canada.ca/climatemodeldata/canrcm/CanRCM4/>). The data from  
865 the 0.44° CanRCM4 large ensemble are available from Environment and Climate Change  
866 Canada (<https://open.canada.ca/data/en/dataset/83aa1b18-6616-405e-9bce-af7ef8c2031c>).  
867

868 NetCDF files of simulated daily streamflow from the historical (1986-2005) and the two future  
869 scenarios (RCP 4.5 and 8.5, 2081-2100) at 0.5° resolution are available on Zenodo for the entire  
870 North American domain of CanRCM4 (doi:10.5281/zenodo.12775139,  
871 <https://zenodo.org/records/12775139>). These streamflow data correspond to the runoff from  
872 the 0.22° simulations.

873

## 874 **Author contributions**

875  
876 VKA designed the study and wrote the majority of the manuscript. AL implemented river  
877 routing to operate at 0.5° resolution and performed all the simulations. RS and AL contributed  
878 to the manuscript text. RS also performed a literature review of existing studies that focus on

879 the impact of climate change on Canadian rivers.

880

### 881 **Competing interests**

882

883 The authors declare that they have no competing interests.

884

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