

How much do we know about snow: past, present and future?

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With major contributions from:

Lawrence Mudryk (ECCC)

Paul Kushner (UToronto)

And colleagues from the CanSISE network.

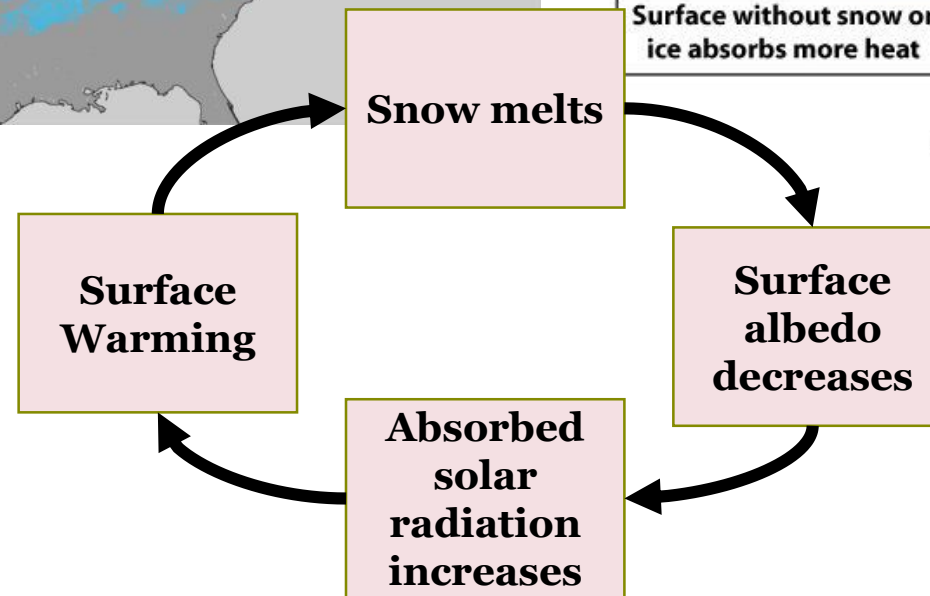
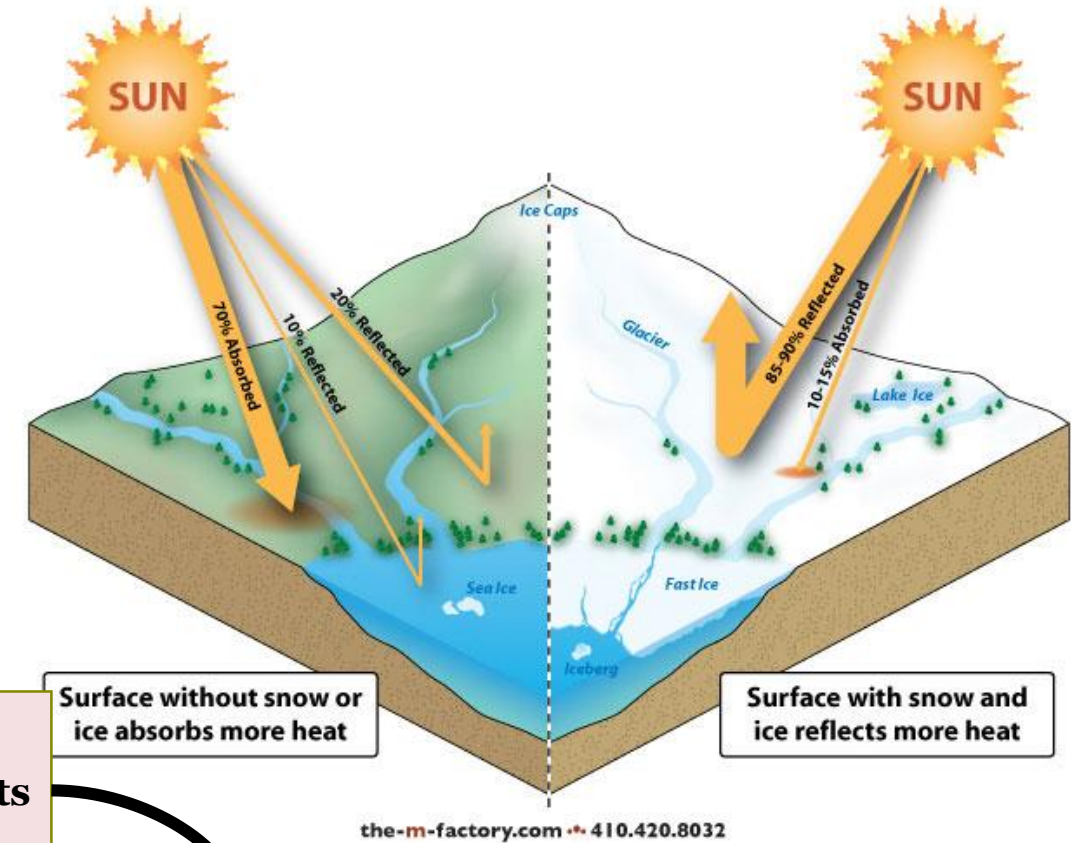
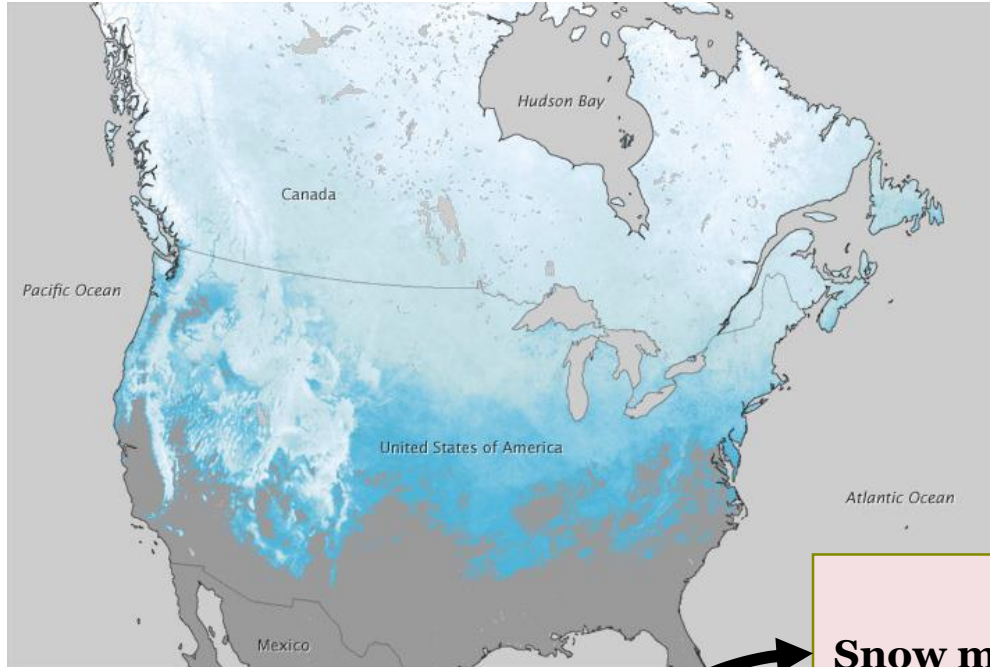


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Importance of snow for climate



Fletcher: How much do we know about snow?

Importance of snowmelt for hydrology



Abstract

A system for operational snowmelt runoff prediction is described as it is presently being implemented for the Rhein-Felsberg basin of the river Rhein in Switzerland. It incorporates a hydrological snow melt model driven by remote sensing data, synoptic measurements and output of a meteorological forecast model. The system components are discussed and first test results are demonstrated.

plied. The SRM is a conceptual degree-day model, based on daily input of air temperature, precipitation and snow cover extent. The latter is derived from satellite data, which enables near real-time assessment of the snow covered area (Baumgartner et al. 1994).

This paper describes the concept and first tests of an operational snowmelt forecast system as it is presently being implemented for the Swiss test basin Rhein-Felsberg within the frame-work of HydAlp (Hydrology of Alpine and High Latitude Basins). This European re-

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DOI 10.1007/s00382-017-3609-x



Rain-on-snow events over North America based on two Canadian regional climate models

Dae Il Jeong¹ · Laxmi Sushama¹



1. Canadian snow: blended observations

15 OCTOBER 2015

MUDRYK ET AL.

8037

Characterization of Northern Hemisphere Snow Water Equivalent Datasets, 1981–2010

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Mudryk, L. R., Derksen, C., Kushner, P. J., & Brown, R. (2015). Characterization of Northern Hemisphere Snow Water Equivalent Datasets, 1981–2010. *Journal of Climate*, 28(20), 8037–8051.



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1. Canadian snow: blended observations

- For model validation and forcing, there is a need for reliable gridded SWE data
- L. Mudryk (ECCC) led a rigorous study of satellite/reanalysis SWE products
- > CanSISE daily NH gridded ($1^\circ \times 1^\circ$) SWE product 1981-2010

Dataset	Abbreviation	Snow scheme	Land model	Forcing data	Resolution	Reference
GlobSnow	GS	Satellite passive microwave + in situ ^a			25 km	Takala et al. (2011)
ERA-Interim/Land	E	Simple	HTESSEL	ERA-Interim	$3/4^\circ \times 3/4^\circ$	Balsamo et al. (2015)
MERRA	M	Intermediate	Catchment	MERRA	$1/2^\circ \times 2/3^\circ$	Rienecker et al. (2011)
Crocus	C	Complex	ISBA	ERA-Interim	$1^\circ \times 1^\circ$	Brun et al. (2013)
GLDAS-2	G2	Simple	Noah 3.3	Princeton Meteorological	$1^\circ \times 1^\circ$	Rodell et al. (2004)
GLDAS-1 ^b	G1n	Simple	Noah 2.7	GDAS + CMAP	$1^\circ \times 1^\circ$	Rodell et al. (2004)
	G1m	Simple	Mosaic			
	G1v	Intermediate	VIC			
	G1c	Intermediate	CLM			
Canadian Meteorological Centre ^b	CMC	Simple + in situ ^c		Global Environmental Multiscale Model (GEM)	35 km	Brasnett (1999); Brown and Brasnett (2010)
MERRA-Land ^b	ML	Intermediate	Catchment	MERRA	$1^\circ \times 1^\circ$	Reichle et al. (2011)

^a GlobSnow is based on combined information from satellite passive microwave retrievals and in situ observations from weather stations. See text for details.

^b Established SWE datasets that meet the NH domain criteria but which contain temporal discontinuities, as analyzed in section 2c, that compromise their use in the multidataset mean.

^c CMC computes snow depths based on combined information from in situ observations and a simple snow scheme, driven by temperature and precipitation from GEM. Depths are converted to SWE using climatological snow density information.

Data available at: <https://nsidc.org/data/nsidc-0668>

2. Trends in Canadian snow (and sea ice) 1981-2015

The Cryosphere, 12, 1157–1176, 2018

<https://doi.org/10.5194/tc-12-1157-2018>

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Canadian snow and sea ice: historical trends and projections

**Lawrence R. Mudryk¹, Chris Derksen¹, Stephen Howell¹, Fred Laliberté¹, Chad Thackeray²,
Reinel Sospedra-Alfonso³, Vincent Vionnet⁴, Paul J. Kushner⁵, and Ross Brown⁶**

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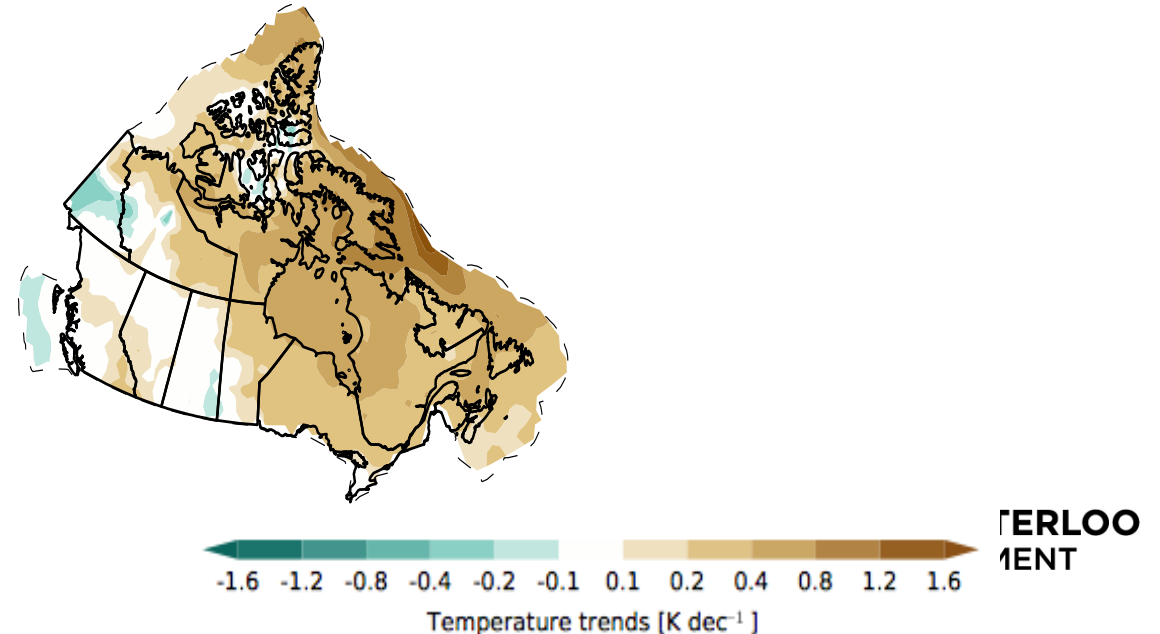
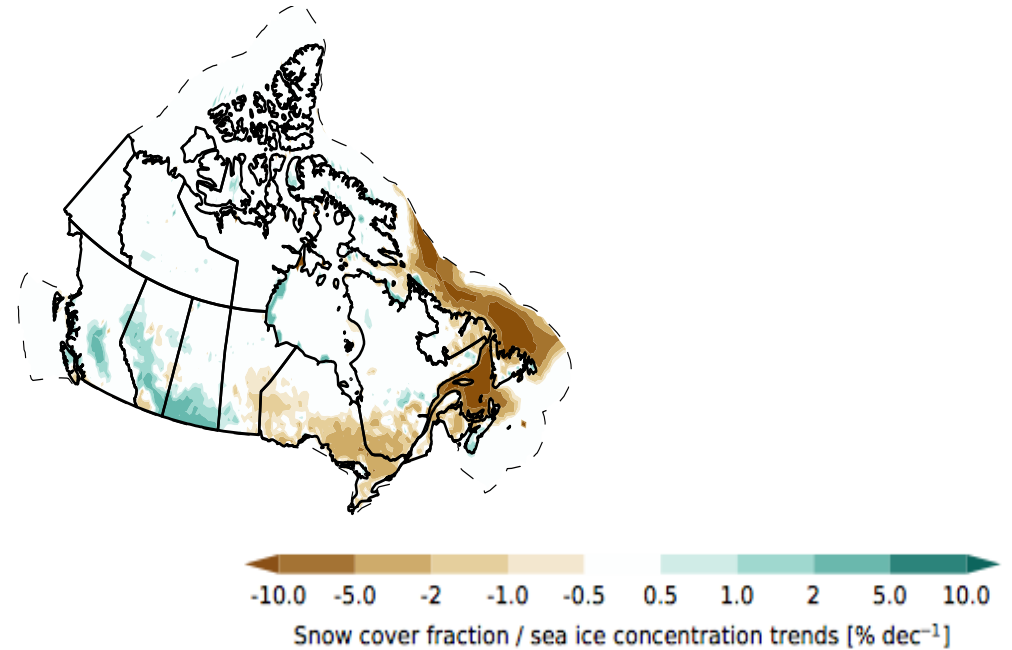
⁵Department of Physics, University of Toronto, Toronto, Canada

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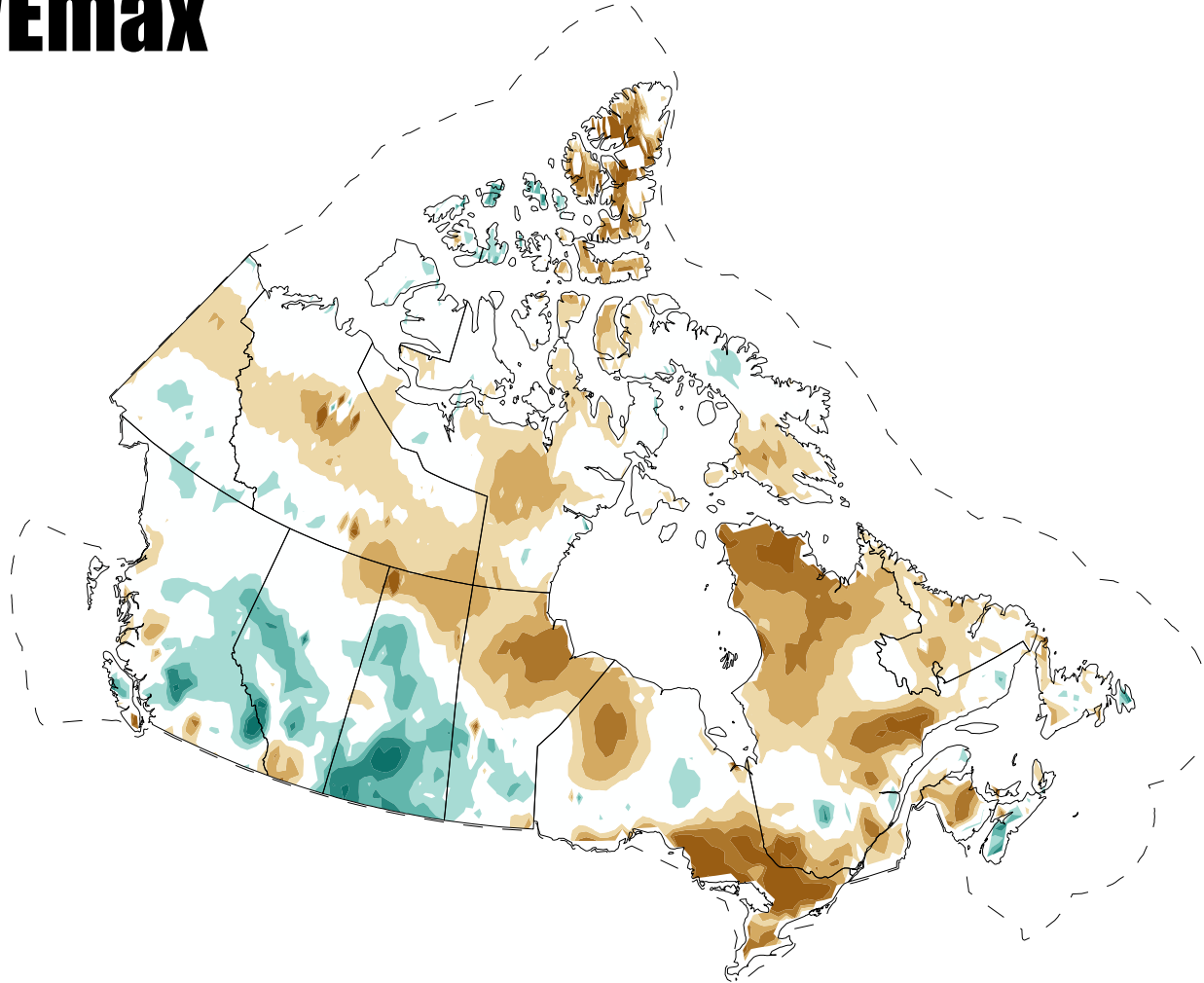


2. Trends in Canadian snow (and sea ice) 1981-2015

- Wintertime snow cover has declined by 5-10% over eastern Canada
- Springtime declines are 5-10% across the Arctic/subarctic
- Snow gains in western Canada
- Spatial pattern of snow trends is closely tied to temperature trends

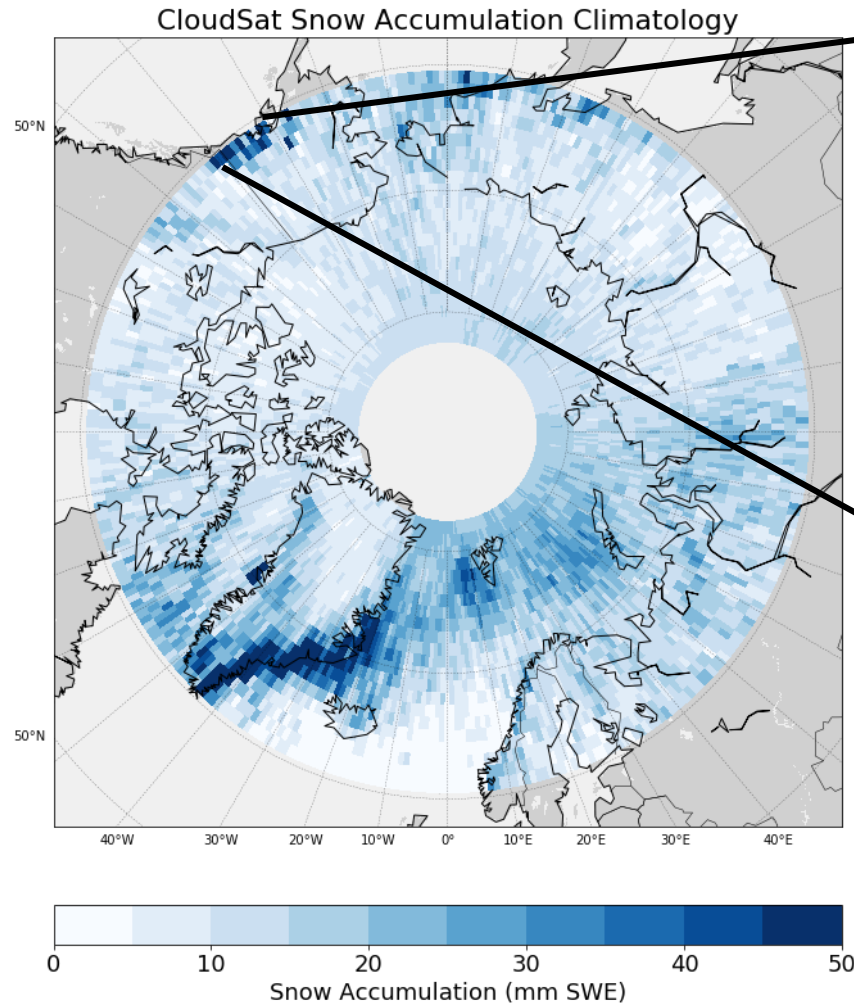


2. Trends in SWE_{max}

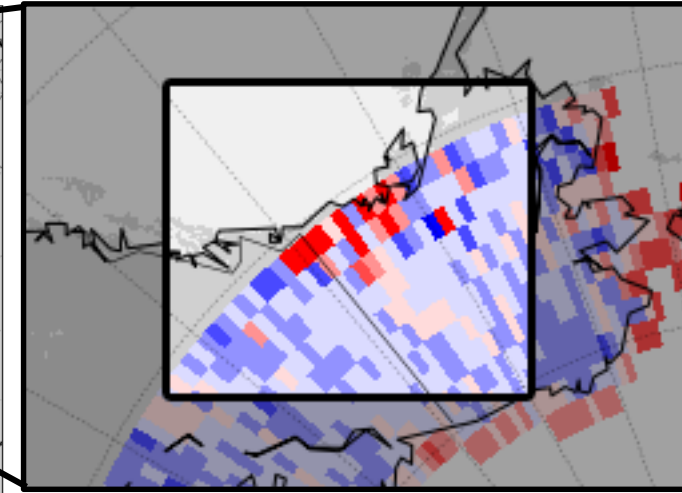


2b. Post-CanSISE re-examination of blended snow data

- CSA-funded project attempting to use space-based retrievals of surface snow to validate gridded products
- The CloudSat-CPR instrument provides a very useful independent look at surface snow climatology
- Highlights potential problems in gridded data over high topography



Bias:



Unpublished analysis by **Fraser King**, MSc student, University of Waterloo.



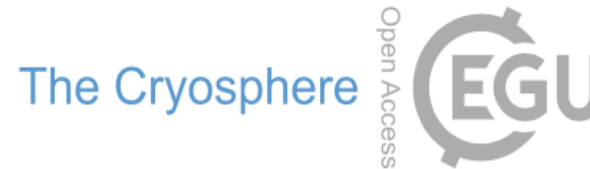
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3. Canadian snow (and sea ice): evaluation of models

The Cryosphere, 12, 1137–1156, 2018

<https://doi.org/10.5194/tc-12-1137-2018>

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Canadian snow and sea ice: assessment of snow, sea ice, and related climate processes in Canada's Earth system model and climate-prediction system

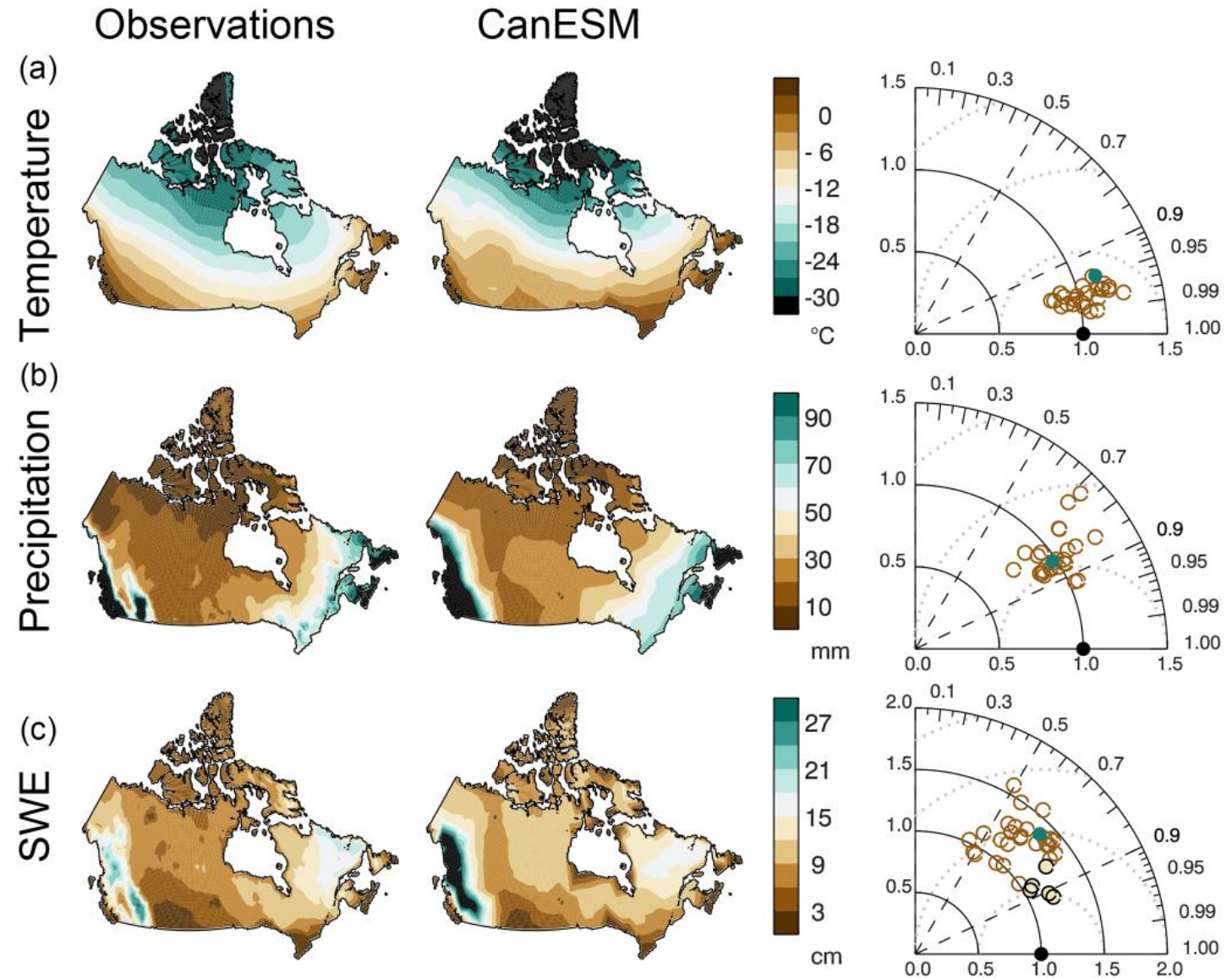
Paul J. Kushner¹, Lawrence R. Mudryk², William Merryfield², Jaison T. Ambadan³, Aaron Berg³, Adéline Bichet⁴, Ross Brown², Chris Derksen², Stephen J. Déry⁵, Arlan Dirkson⁶, Greg Flato², Christopher G. Fletcher⁷, John C. Fyfe², Nathan Gillett², Christian Haas^{8,9}, Stephen Howell², Frédéric Laliberté², Kelly McCusker¹⁰, Michael Sigmond², Reinel Sospedra-Alfonso², Neil F. Tandon², Chad Thackeray⁷, Bruno Tremblay¹¹, and Francis W. Zwiers¹²



3. Canadian snow: models vs. observations

- P. Kushner (U Toronto) led an evaluation of the Canadian Earth System Model (CanESM2)
- All models simulate JFM mean temperature pattern very well
- Larger disagreement for precip and SWE, but CanESM2 is among the best models.

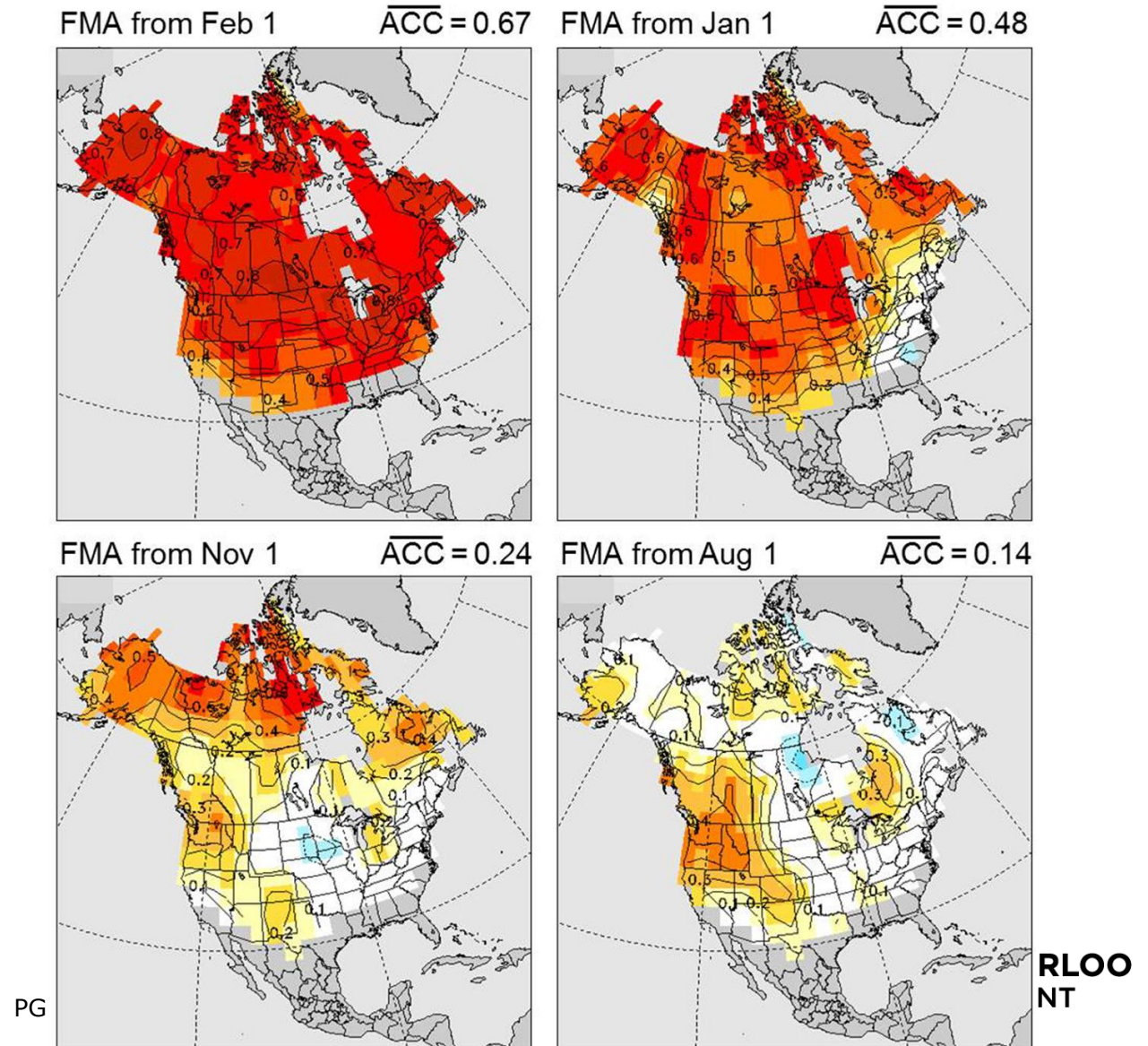
Kushner et al. (2018)



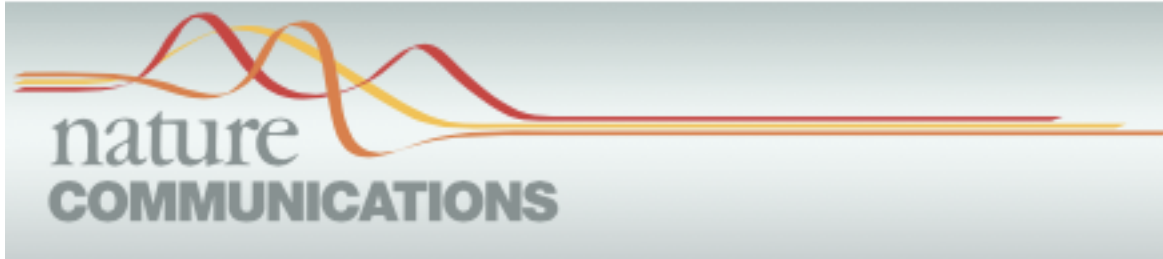
3. Canadian snow: SWE prediction skill

- Colleagues at CCCma (Victoria) and U Guelph showed spring SWE is highly predictable from Nov onwards.
- Highest skill is found over the Canadian Arctic/subarctic, and western cordillera
- Lower skill in central/east
- Improved initialisation of surface snow yields much better forecast skill (not shown).

Kushner et al. (2018)



4. Near-term projections of snowpack loss (western US)



ARTICLE

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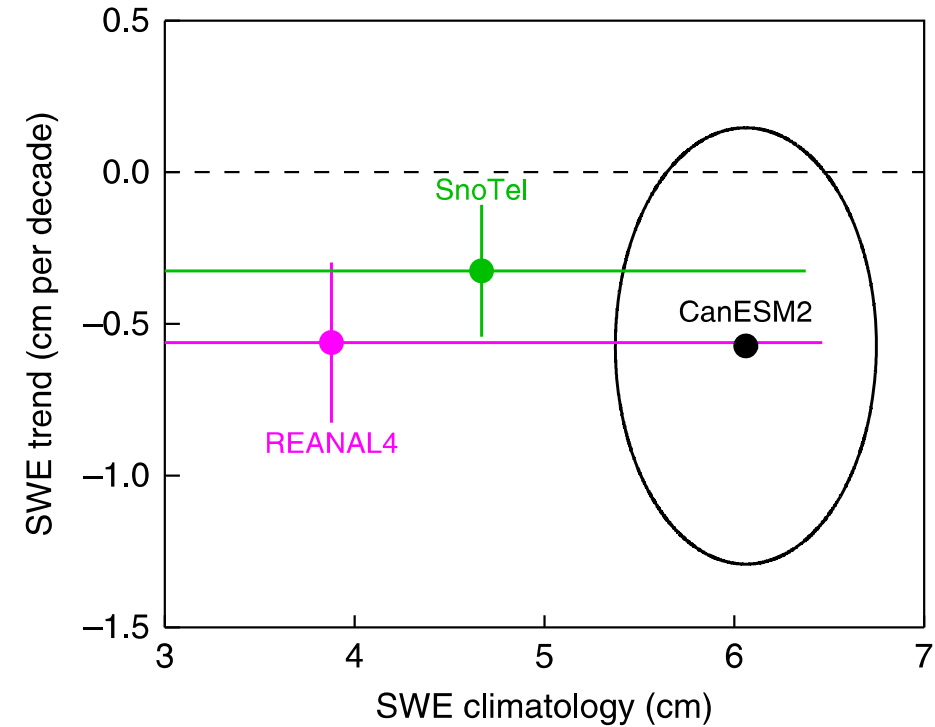
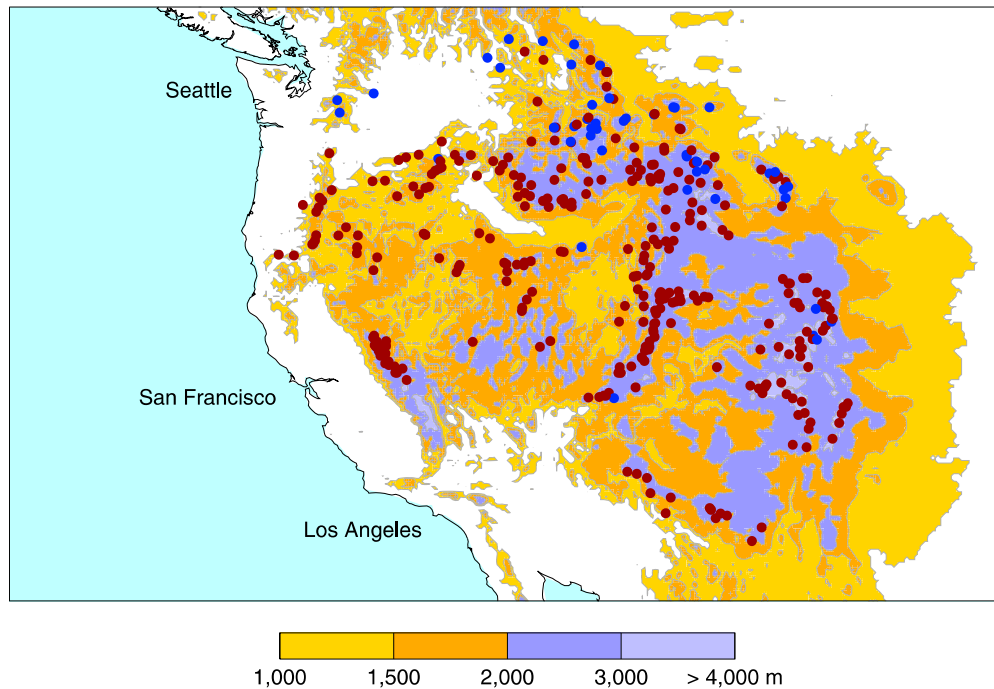
DOI: 10.1038/ncomms14996

OPEN

Large near-term projected snowpack loss over the western United States

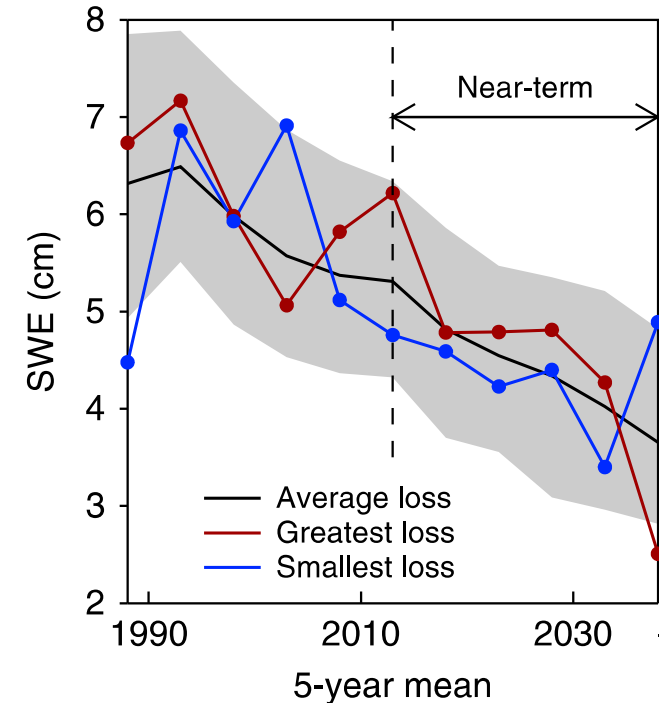
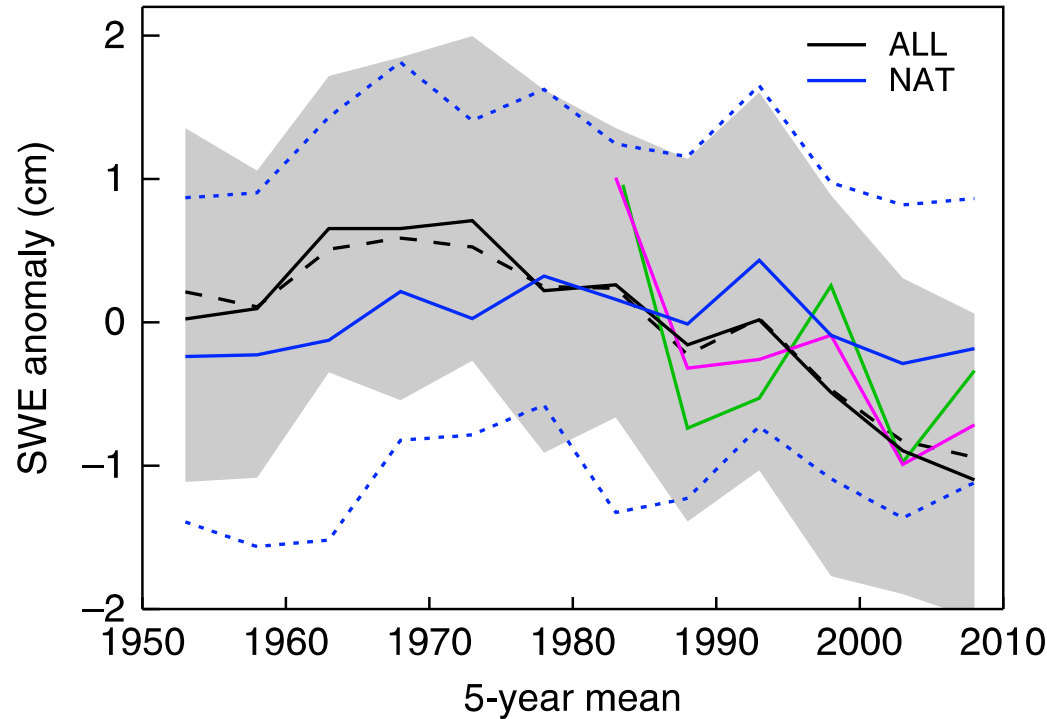
John C. Fyfe¹, Chris Derksen², Lawrence Mudryk², Gregory M. Flato¹, Benjamin D. Santer³, Neil C. Swart¹, Noah P. Molotch^{4,5}, Xuebin Zhang², Hui Wan², Vivek K. Arora¹, John Scinocca¹ & Yanjun Jiao¹

4. Near-term projections of snowpack loss (western US)



- J. Fyfe (CCCma) led a study showing 10-20% losses in snowpack since 1980
- CanESM2 model is in good agreement with gridded and in situ observations

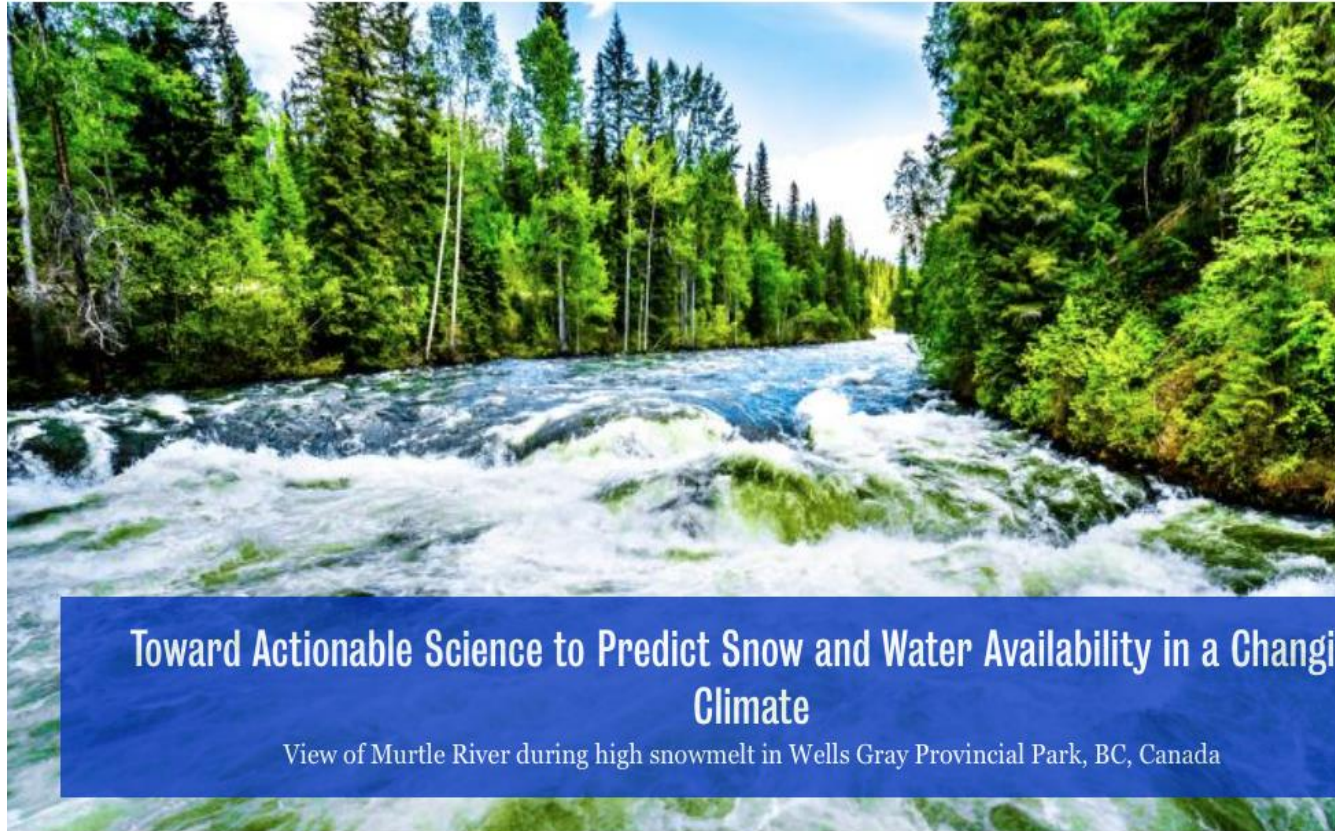
4. Near-term projections of snowpack loss (western US)



- Observed trends are consistent with historical forcings, but not with “natural only”
- Projections show GHG forcing associated with continued rapid snowpack loss
- However, internal climate variability has a large influence on near-term trends.



SNOWMELT WORKSHOP 2018



Toward Actionable Science to Predict Snow and Water Availability in a Changing Climate

View of Murtle River during high snowmelt in Wells Gray Provincial Park, BC, Canada

[Home](#) [Event Details](#) [Travel and Accommodation](#) [About](#)

<https://uwaterloo.ca/scholar/bmlyons>

Keynote Speakers:

- Prof Paul Kushner (UToronto)
- Dr Andre Erler (Aquanty Inc)
- Ms Chelsea Mottishaw (City of Dawson Creek)
- Dr Hank Venema (Strategic Community Consulting)
- Dr Stephanie Shifflett (GRCA)

Number of attendees by sector:

- | | |
|--------------|---|
| • Academia | 7 |
| • Government | 6 |
| • NGO | 5 |
| • Industry | 5 |

Workshop Coordinator: Blaine Lyons, bmlyons@uwaterloo.ca

The end.

`chris.fletcher@uwaterloo.ca`

`https://uwaterloo.ca/scholar/c5fletch`



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2. Physical controls on SWE_{max}

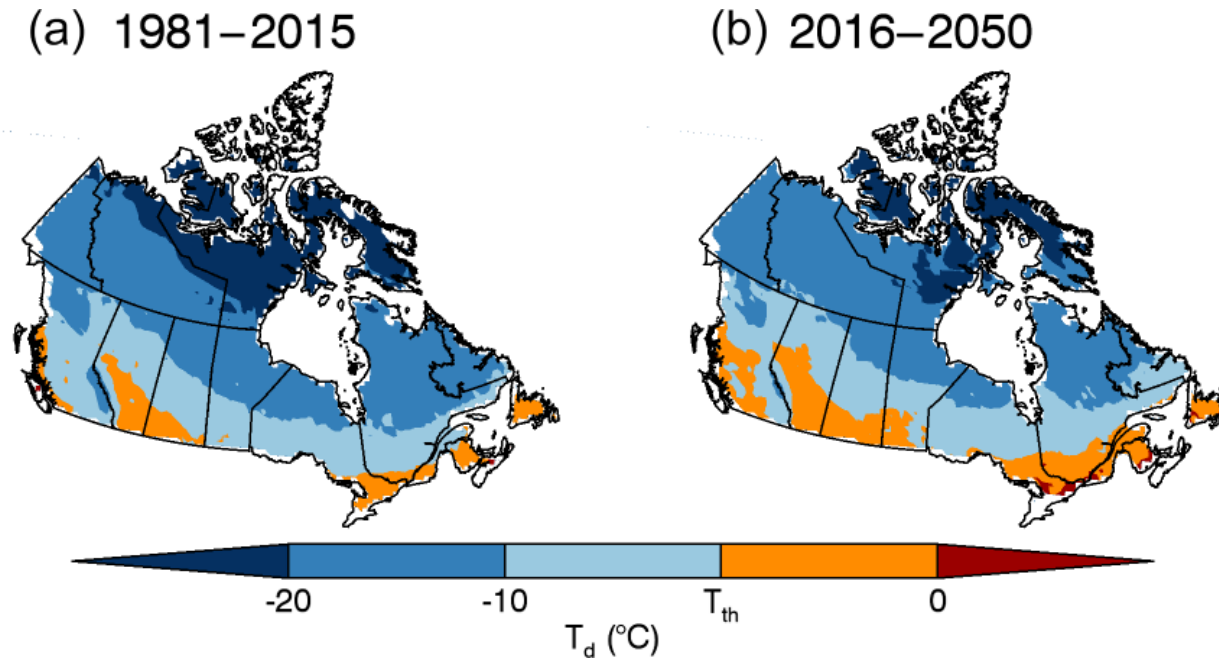


Figure 12. Temperature and precipitation controls on March snow water equivalent for historical period (a) and projections to 2050 (b). For the data presented here $T_{th} = -5.4^\circ\text{C}$ (more generally $T_{th} = -5 \pm 1^\circ\text{C}$). Regions with $T_d < T_{th}$ (blue) have March SWE dominated by precipitation variability while regions with $T_d > T_{th}$ (orange) have March SWE dominated by temperature variability.