



# Toward user-relevant monthly to seasonal forecasts of **Arctic sea ice: The FRAMS project**

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#### **About FRAMS**

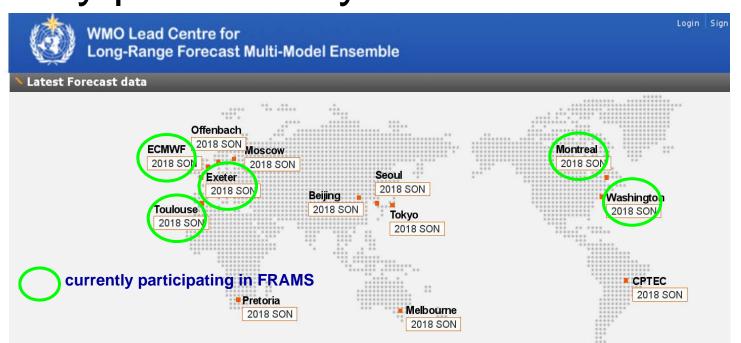
<u>Forecasting Regional Arctic Sea Ice from a Month to Seasons</u> (FRAMS) is a three-year project funded by Canada's Marine Environmental Observation Prediction and Response (MEOPAR) network. It is endorsed by the Year of Polar Prediction (YOPP).

#### **Objectives**

- 1) Advancing the science of multi-model sea ice forecasting on time scales of a month to seasons
- 2) Developing Arctic sea ice forecast products and services for the new WMO Arctic Polar Regional Climate Centre (ArcRCC)
- 3) Identifying physical processes and aspects of initial states that

### WMO seasonal forecasting and the ArcRCC

- WMO seasonal forecasts currently provided by 13 Global Producing Centres (GPCs) WMO Lead Centre for Long-Range Forecast Multi-Model Ensemble
- Of the GPCs whose models have interactive sea ice, 5 are currently participating in FRAMS  $\longrightarrow$



The ArcRCC has 3 nodes, in Norway, Russia and Canada. The Canadian node in Montreal is tasked with providing forecast information, including for sea ice.

#### FRAMS analysis component

enable sea ice to be skillfully predicted

4) Co-developing, with the Canadian Ice Service and end users in the Arctic marine transportation sector, sea ice forecast products that are useful for decision making

## FRAMS forecasting component

The forecasting component of FRAMS is assembling data from the following models to support **research and development of** products and services:

label	name	centre	sea ice component, properties, rheology	max resolution / forecast range
M1	CanCM3/4	ECCC/MSC	concentration/thickness, cavitating fluid	≈200 km / 12mon
M2	GEM-NEMO	ECCC/MSC	CICE, 5 ice categories, EVP	≈ 40 km / 12mon
M3	CFSv2	NOAA (US)	GFDL SIS, 5 ice categories, EVP	≈ 40 km / 9 mon
M4	System 5	Météo France	GELATO, 4 ice categories, EVP	$\approx$ 40 km / 7 mon
M5	GloSea5	Met Office (UK)	CICE, 5 ice categories, EVP	≈ 10 km / 5 mon
M6	SEAS5	ECMWF	CICE, 5 ice categories, EVP	≈ 10 km / 7 mon
M7	En-GIOPS	ECCC/MSC	CICE, 10 ice categories, EVP	≈ 10 km / 1 mon

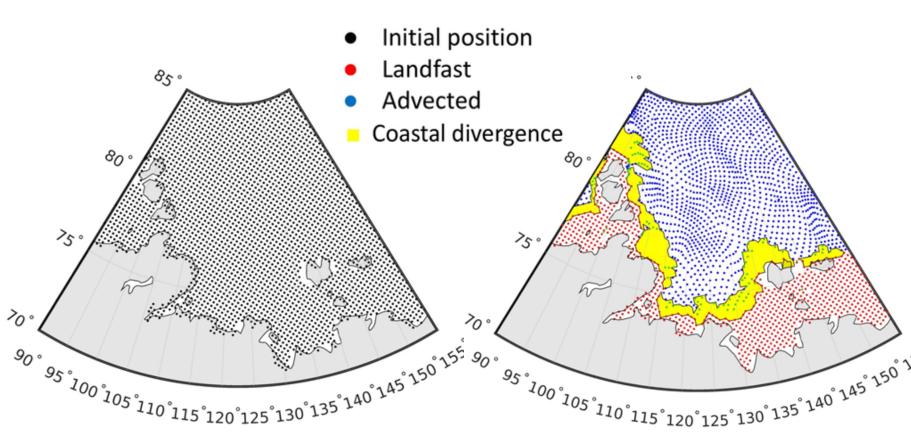
• These models differ in their spatial detail:

#### **Questions asked:**

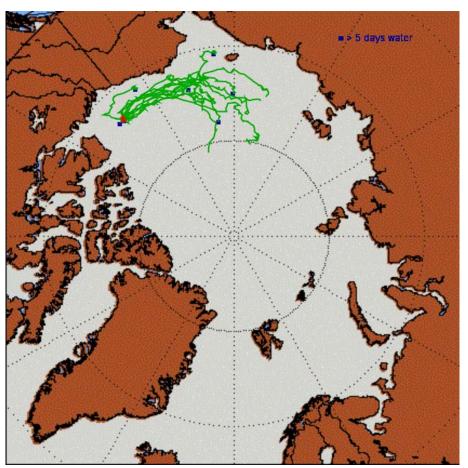
- What are the sources of skill in sea ice forecasts?
- How are models deficient in representing these key processes?
- How do current model forecasts compare to empirical predictions?

# Key tool: Lagrangian Sea Ice Tracking System (LITS)

Follows sea ice trajectories based on Polar Pathfinder ice motions  $\longrightarrow$ (Williams et al. 2016)





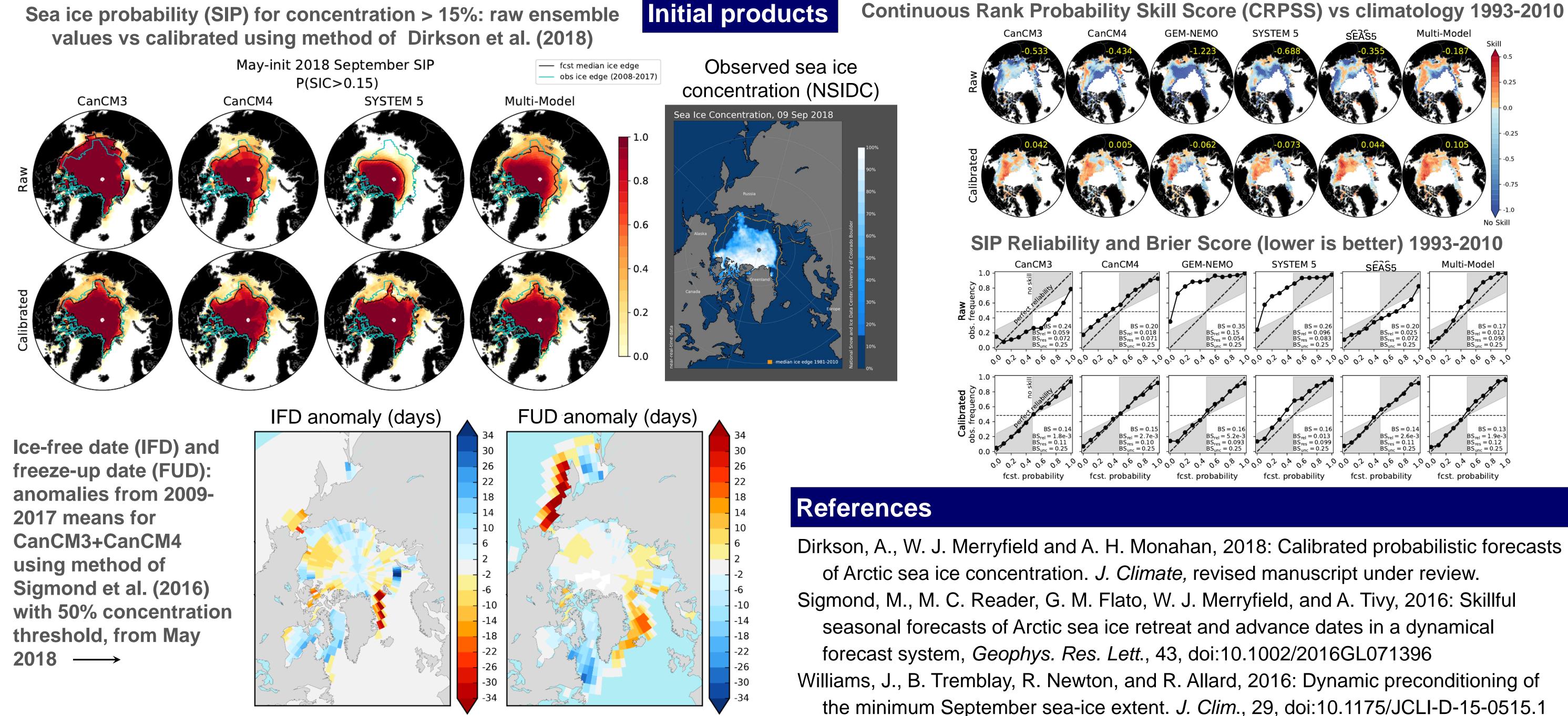


12-month sea ice trajectories from initial location in the Beaufort Sea, from Sep 2000-2013

**Coastal divergence in the Laptev** Sea results in thinner ice, lower September ice extent C. Brunette/McGill



- M1 somewhat skillful despite coarse grid  $\rightarrow$  development platform
- **M2-M4** reflect typical resolution in current sea ice forecast models
- **M5-M7** reflect leading-edge current and emerging capabilities
- Aim is to **co-develop** with end users & Canadian Ice Service forecast products relevant to Arctic navigability, decision making
- Emphasizing communication of forecast uncertainty
- Workshop with end users in May 2018, another in 2020



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Dirkson, A., W. J. Merryfield and A. H. Monahan, 2018: Calibrated probabilistic forecasts