



National Snow and Ice Data Center
Supporting Cryospheric Research Since 1976



Land Data Assimilation for Seasonal Prediction and Beyond: The Arctic Case

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Uncertainty in Numerical Forecasting

(1) Model Structure

- Parameterizations
- Piecing together components
- Numerical methods

(2) Model Forcing

- Spatial & Temporal structure

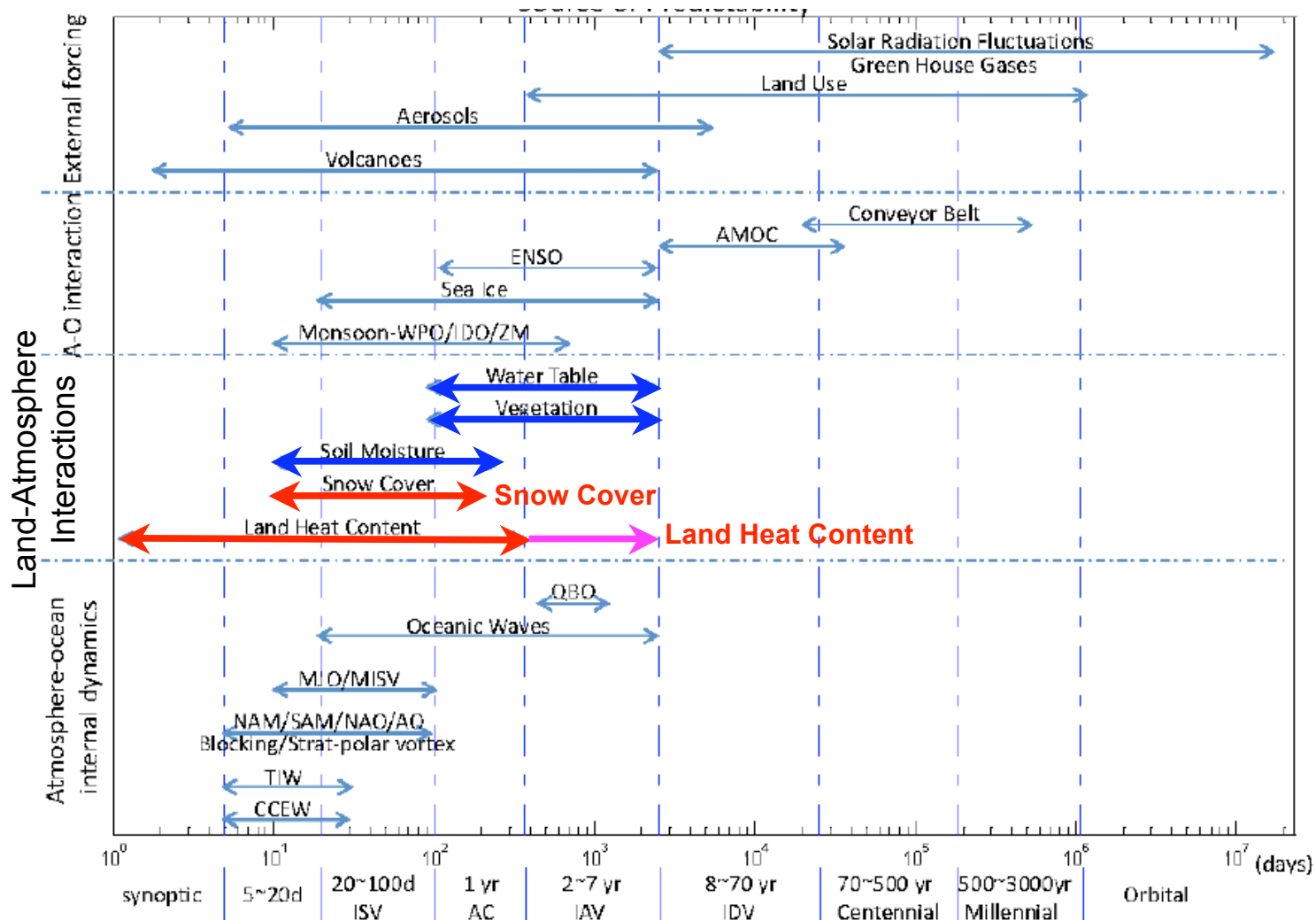
(3) Parameter Data

- Soils & Vegetation, type and distribution

(4) Initial Conditions

- Influences trajectory (forecasting = IVP)

Sources of Predictability



An Assimilation Scheme: The Kalman Filter

1. $\mathbf{X}_t^- = \mathbf{A}\mathbf{X}_{t-1} + \mathbf{B}f_t$

2. $\mathbf{K}_t = \mathbf{P}_t\mathbf{H}^T(\mathbf{H}\mathbf{P}_t\mathbf{H}^T + \mathbf{R})^{-1}$

3. $\mathbf{X}_t^+ = \mathbf{X}_t^- + \mathbf{K}_t(\mathbf{z}_t - \mathbf{H}\mathbf{X}_t^-)$

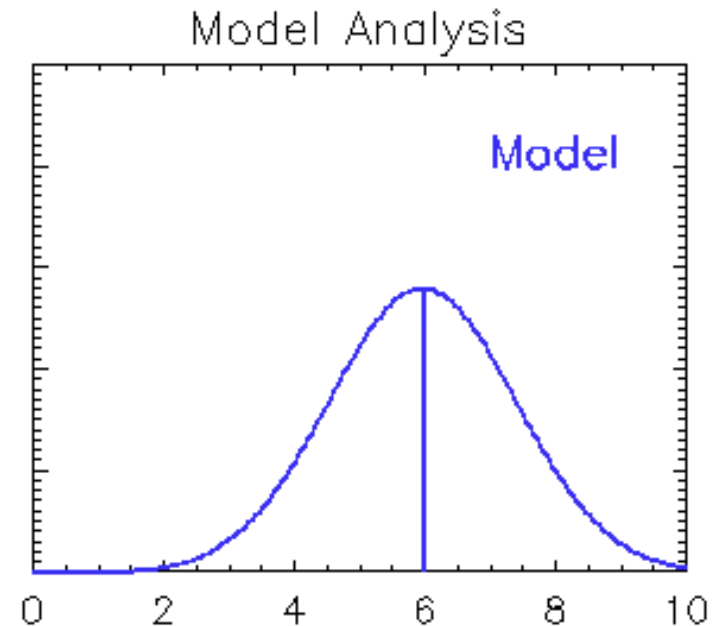
- \mathbf{X} – model state vector
- \mathbf{P} – model error
- \mathbf{H} – measurement model
 - equating quantities
- \mathbf{R} – observational error
 - representativeness error
 - measurement error
- \mathbf{K} – Kalman gain
 - relate obs and states
- \mathbf{z} - observations

Optimized Assimilation : Scalar Example

Our **Model** predicts : $X^- = 6$

Model error variance : $P = \sigma_x^2 = 2$

Assume $H = 1$ (no impact)



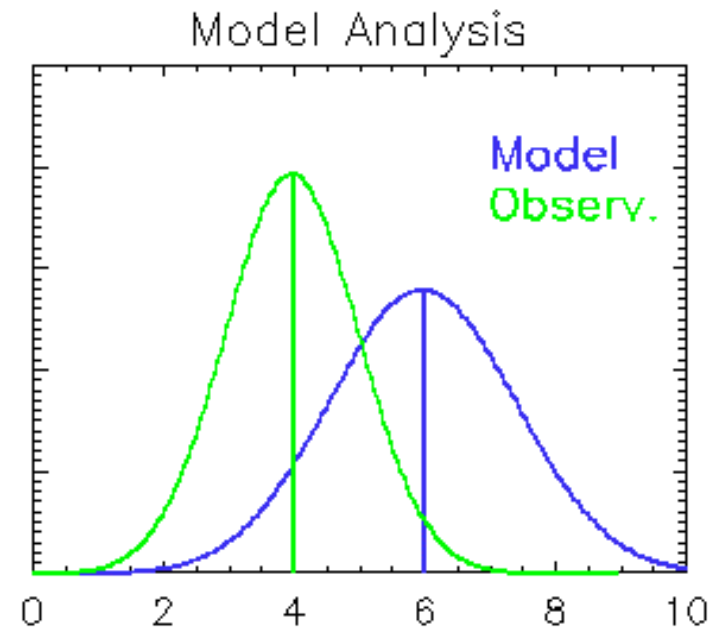
Optimized Assimilation : Scalar Example

Our **Observations** say : $Z = 4$

Obs. error variance : $R = \sigma^2_z = 1$

Gain = $2/(2 + 1) = 0.66$

Analysis = $6 + 0.66(4 - 6)$



Optimized Assimilation : Scalar Example

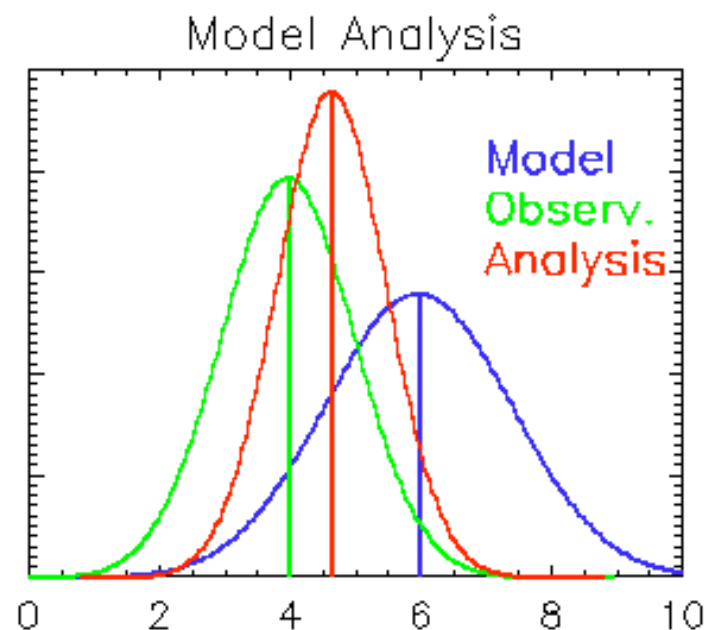
Combined **Model** and **Observations** say :

Our **Analysis** is $X^+ = 4.66$

Analysis variance : $\sigma_a^2 = 0.66$

$$\frac{1}{\sigma_a^2} = \frac{1}{\sigma_x^2} + \frac{1}{\sigma_z^2}$$

Analysis Variance



An Assimilation Scheme: The Kalman Filter

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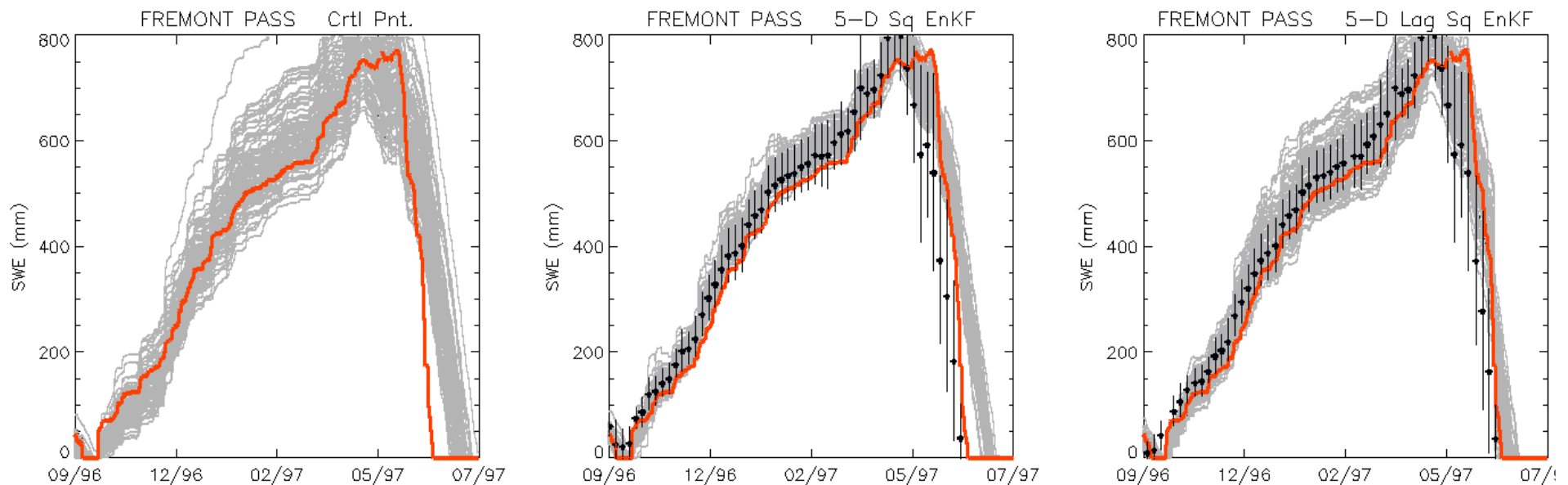
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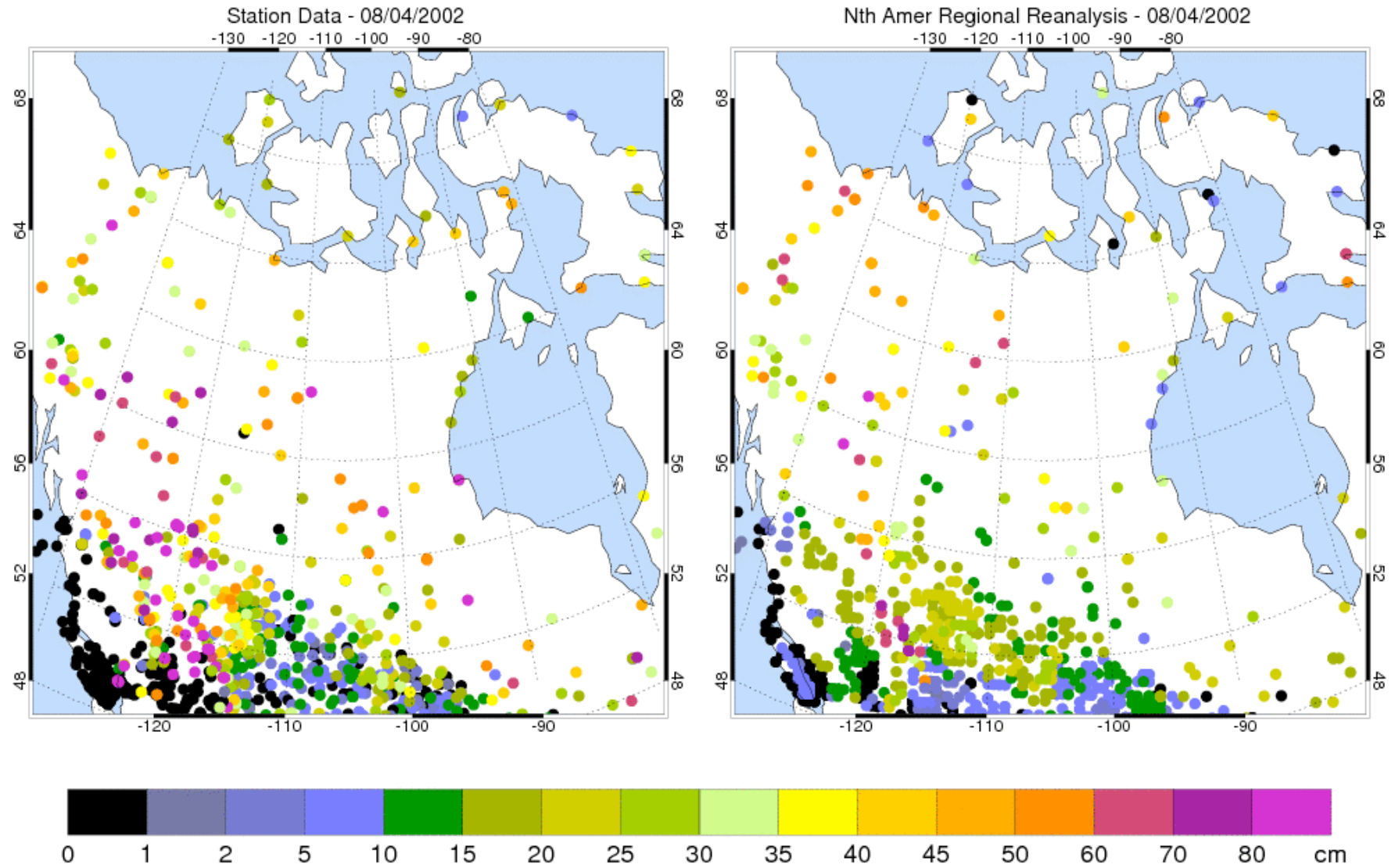
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Station Based Snow Data Assimilation

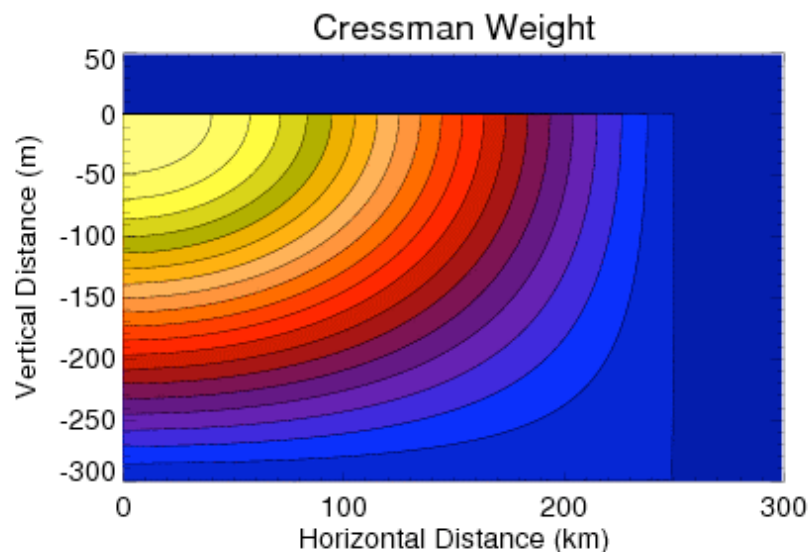
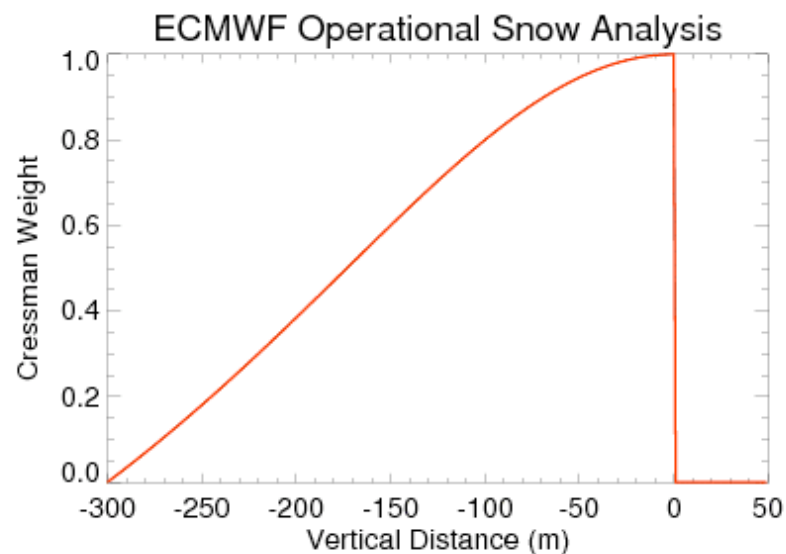
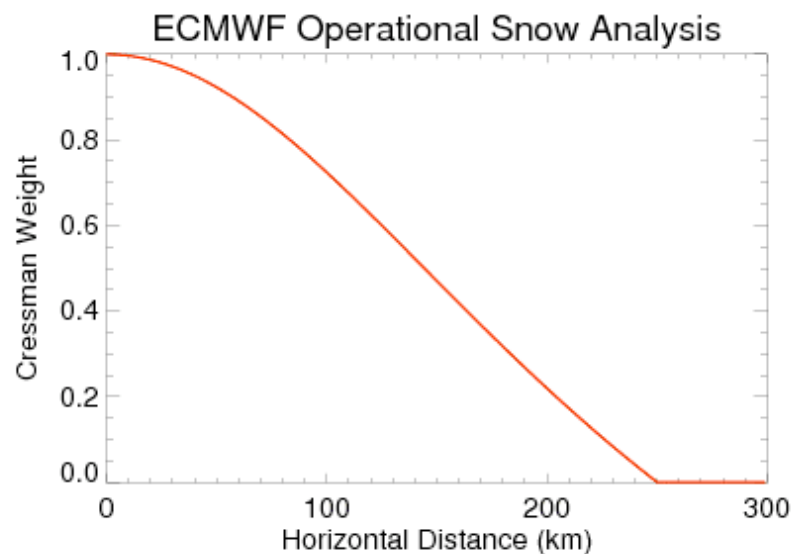
- Good station density
- Unbiased model
- Appropriate error estimation
- Meet assumptions of assimilation scheme
- Successful analysis result



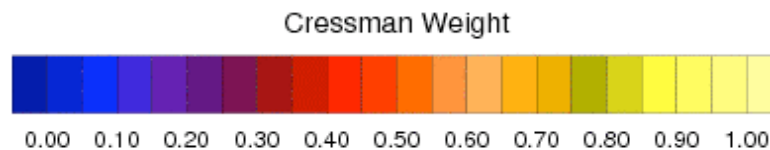
Low Station Distribution in Arctic

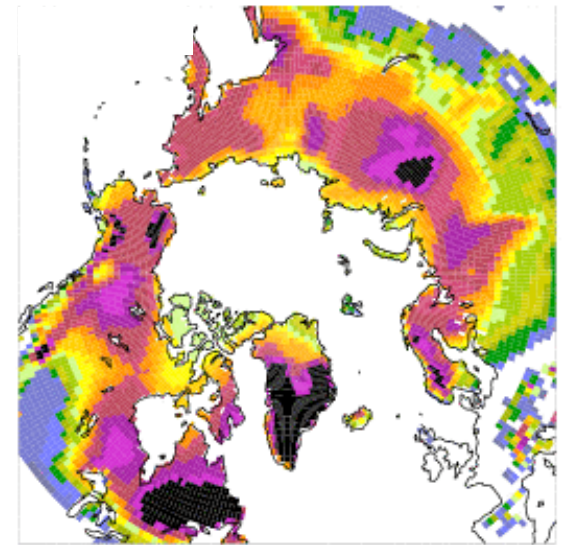
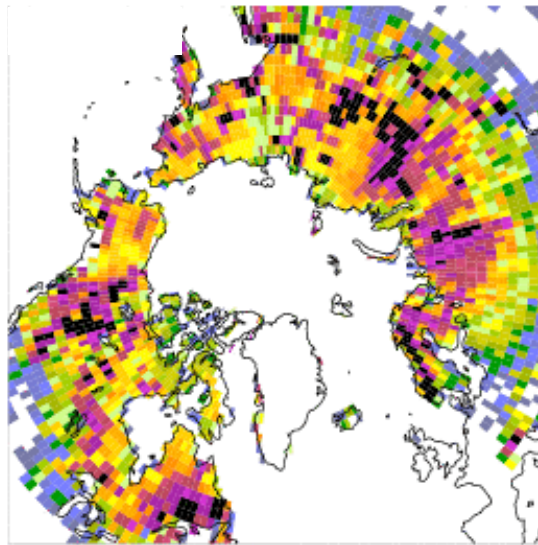
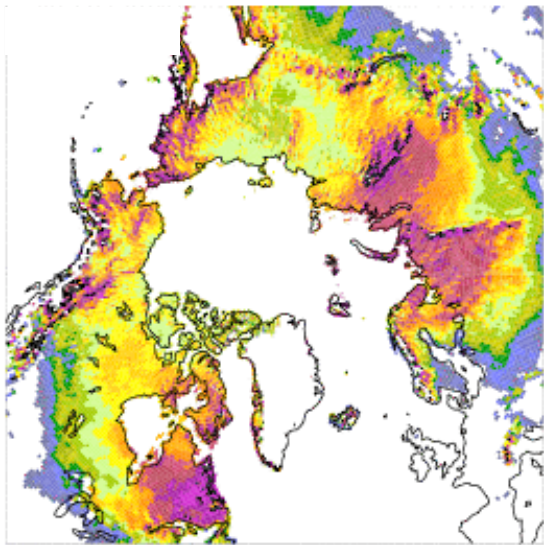


ECMWF Station-based Snow Analysis

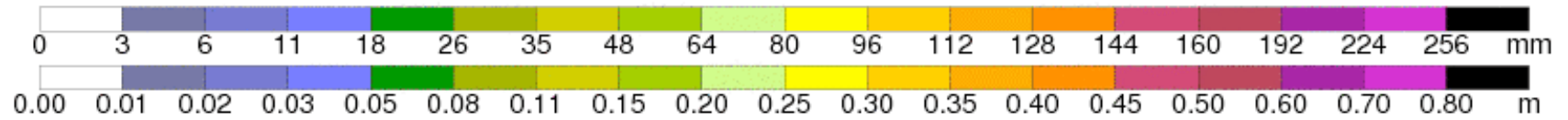
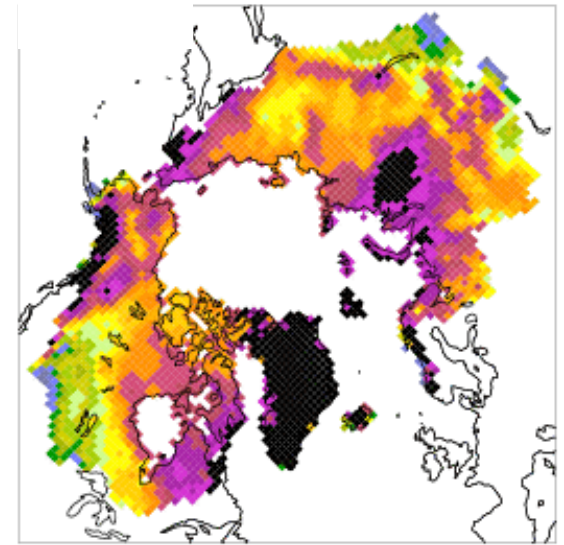
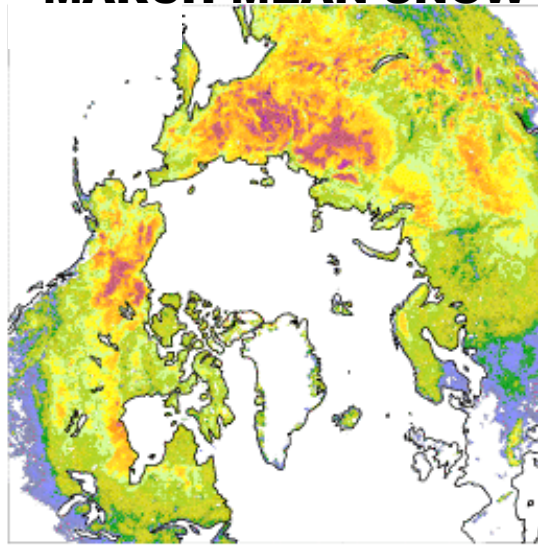
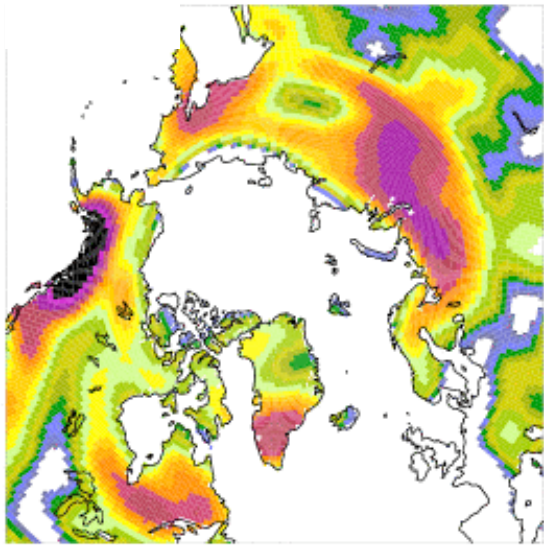


Need two data points for analysis

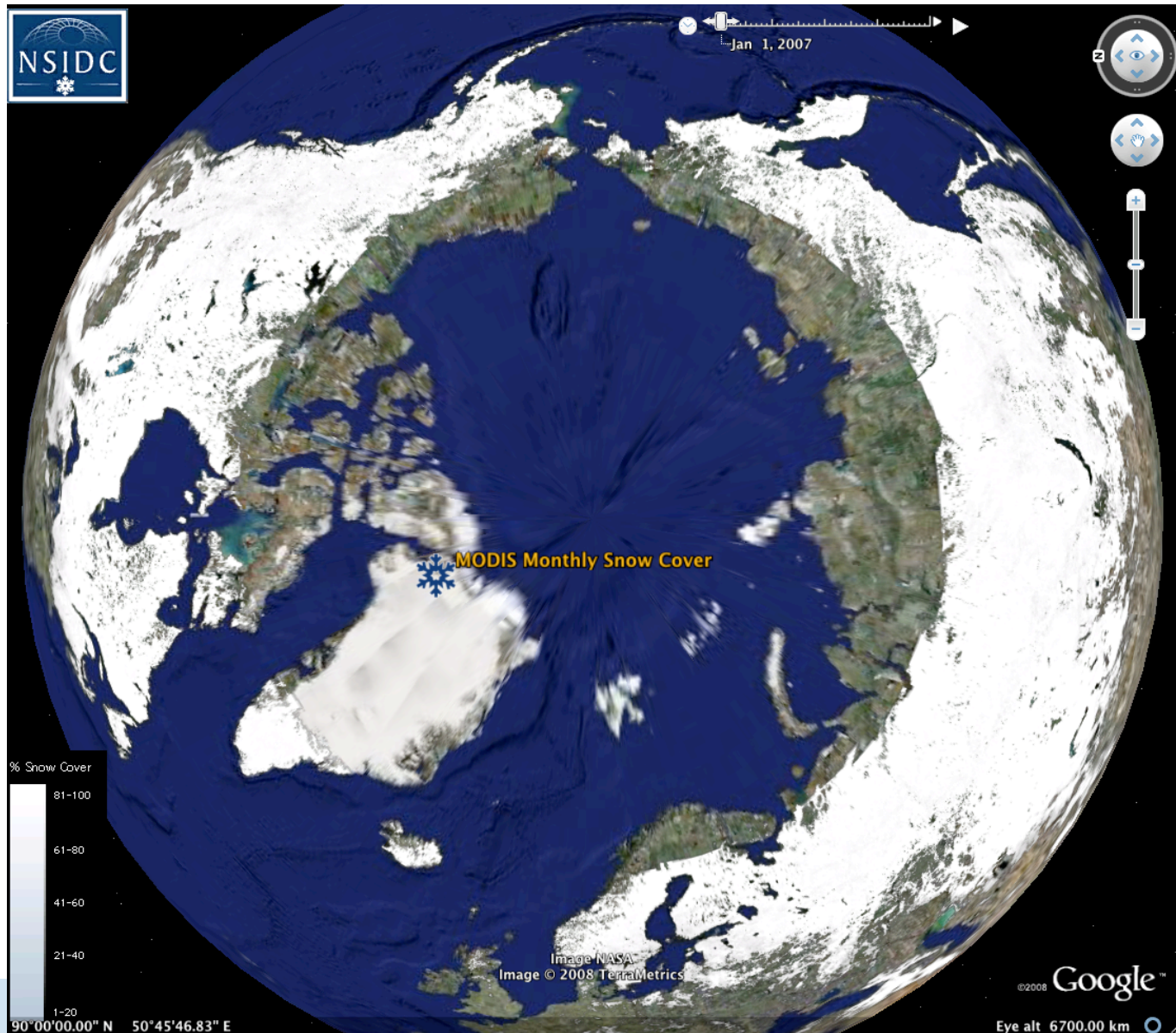




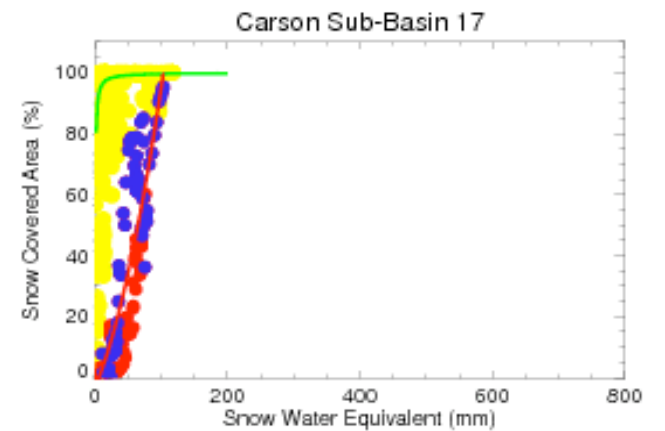
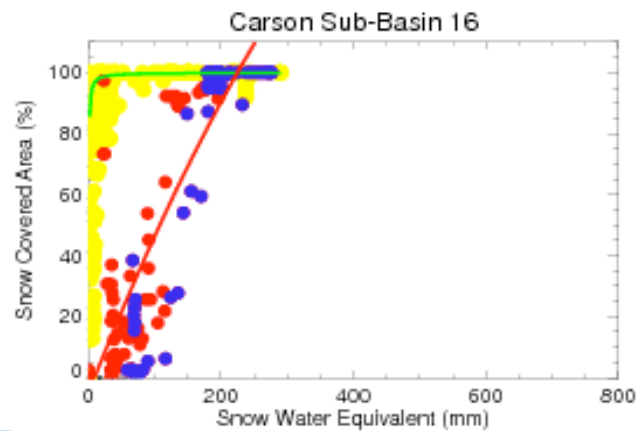
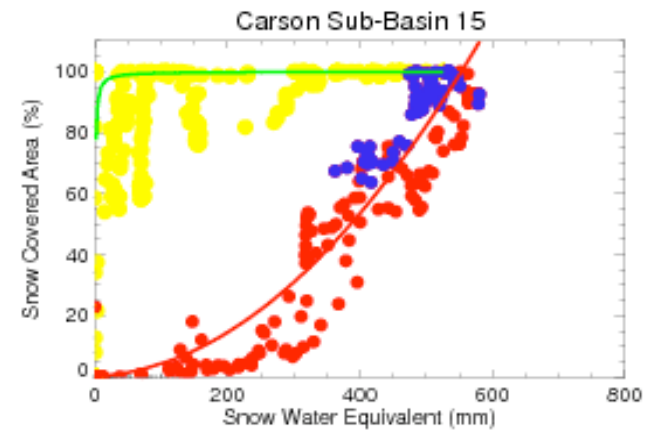
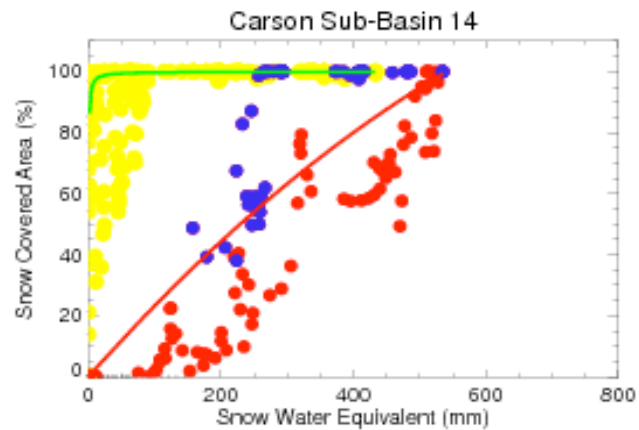
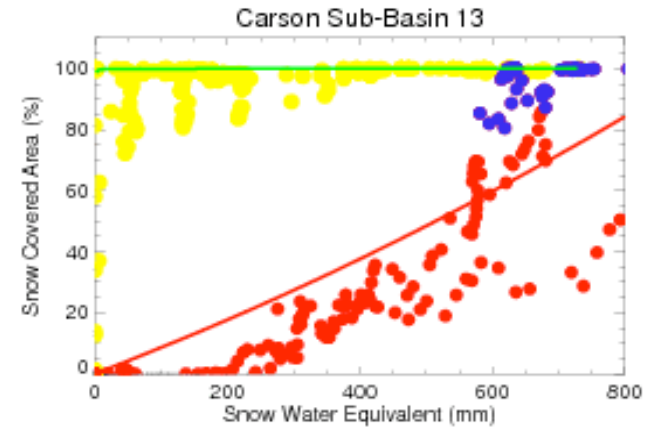
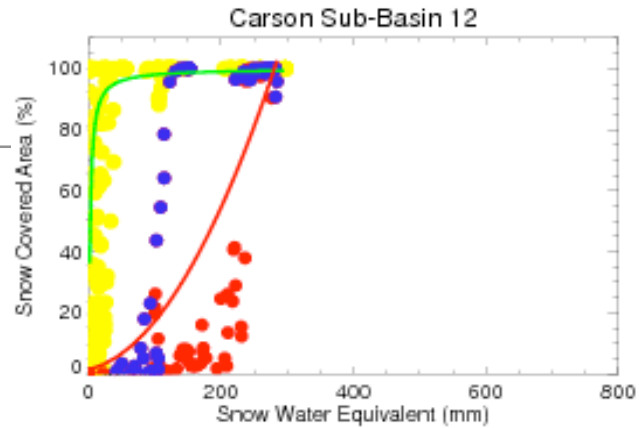
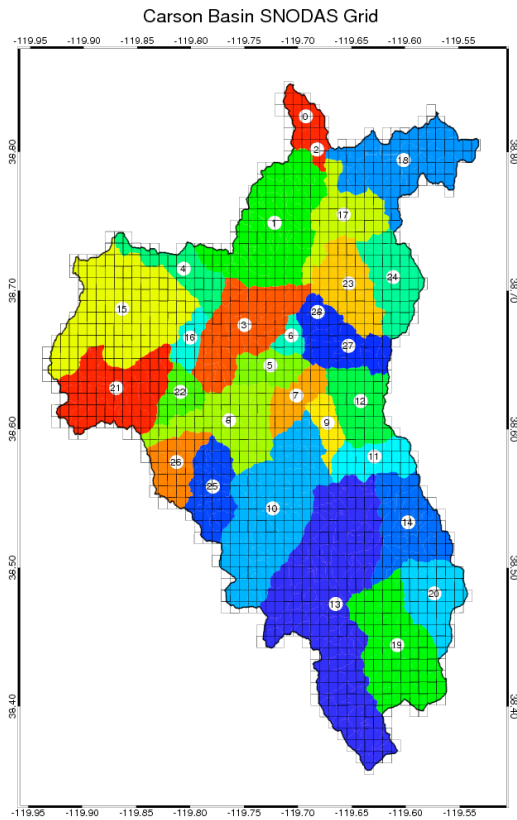
MARCH MEAN SNOW



Snow Extent



MODIS SCA VS SNODAS SWE

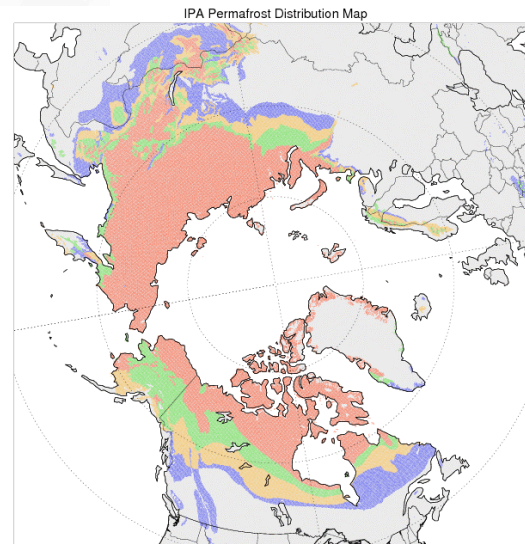


Pan-Arctic Snow Data Prospects

	Z (coverage)	H (Obs to Model)	R (error estimate)
Station Data (Depth & SWE)	Spotty Daily	Point vs Area Depth to SWE	Point vs Grid Box
Passive Microwave (SWE)	Global Daily	Radiance Model	Crystal, Veg, Tmp
Visible – Extent Visible – Volume	Global Daily	1km Pixels SCA → SWE	Ideal Poor
Gravity - GRACE (Mass anomaly)	Global Monthly	Huge Area vs Grid box	SWE vs other water

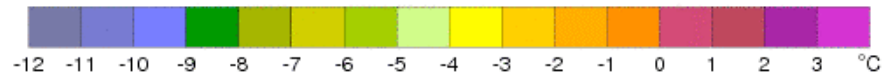
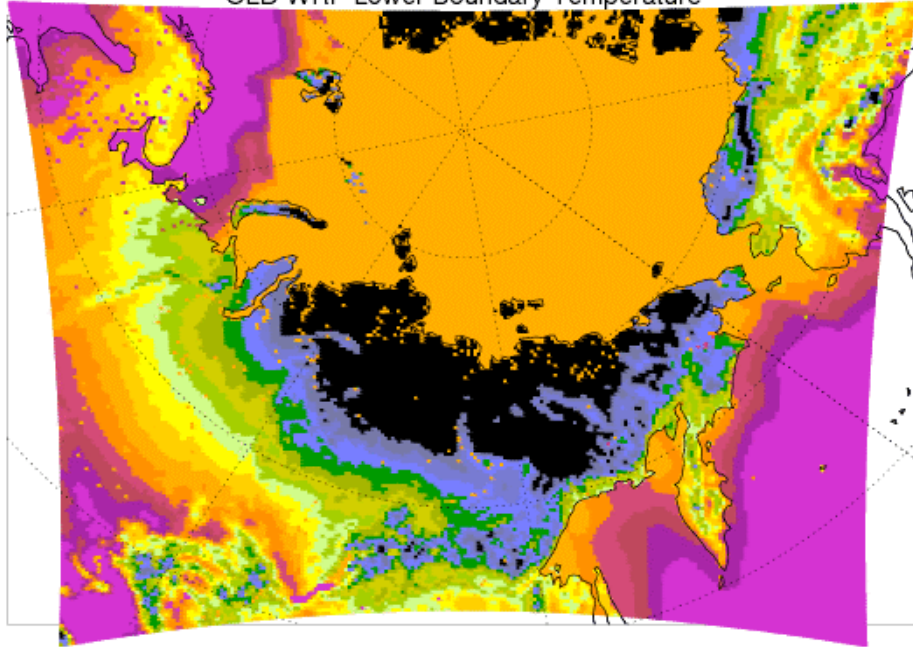
Ground Thermo-Hydro State

- Soil phase change, latent heat store
- Deep soil heat reservoir
- Altered hydrologic regime



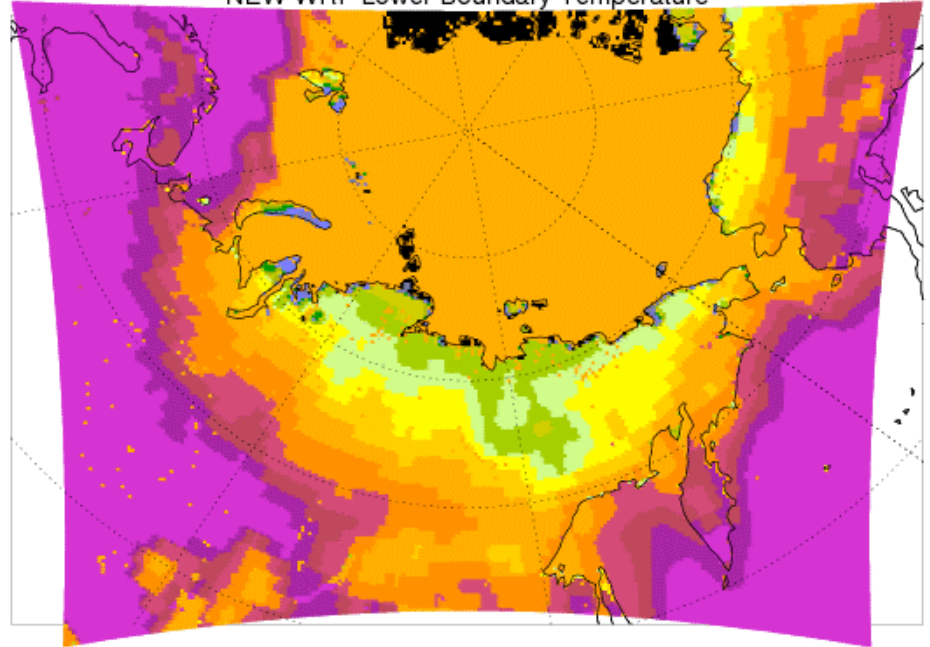
Thermal State – Boundary Conditions

OLD WRF Lower Boundary Temperature



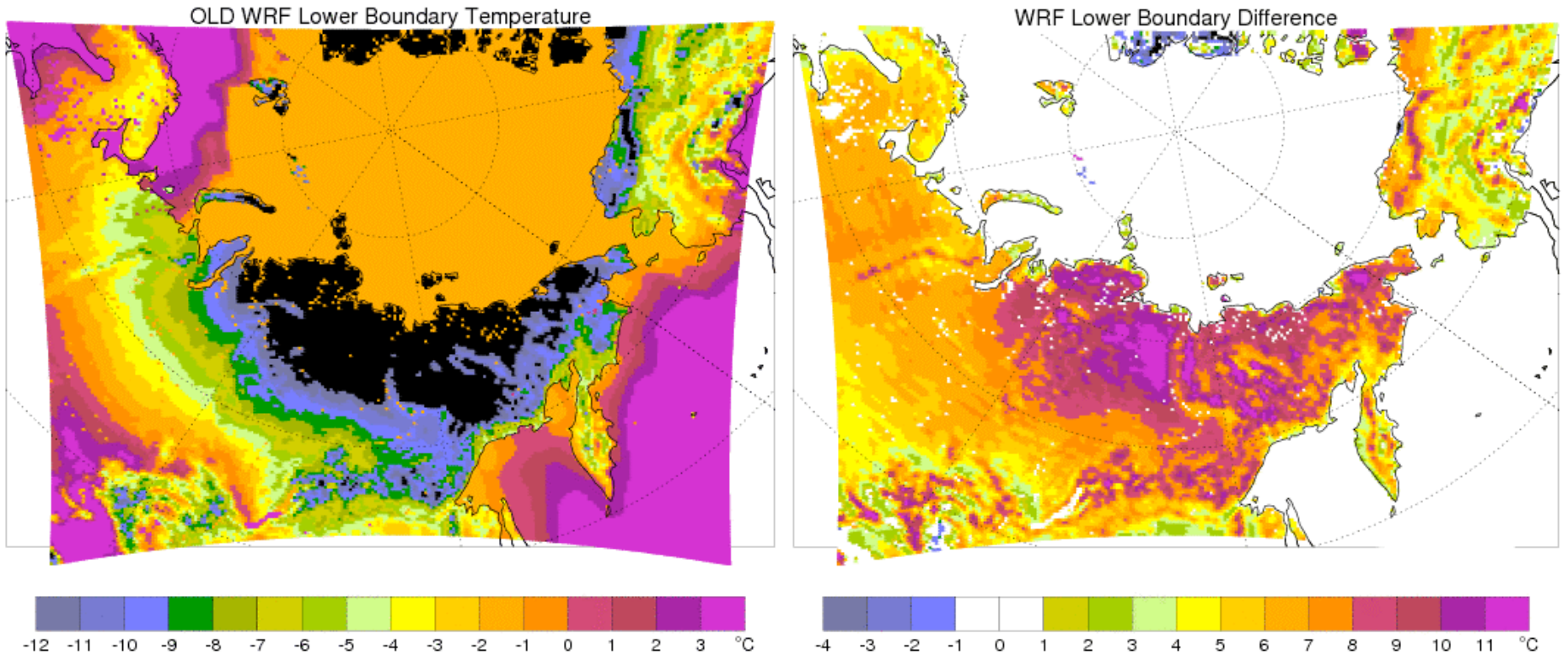
OLD

NEW WRF Lower Boundary Temperature



NEW

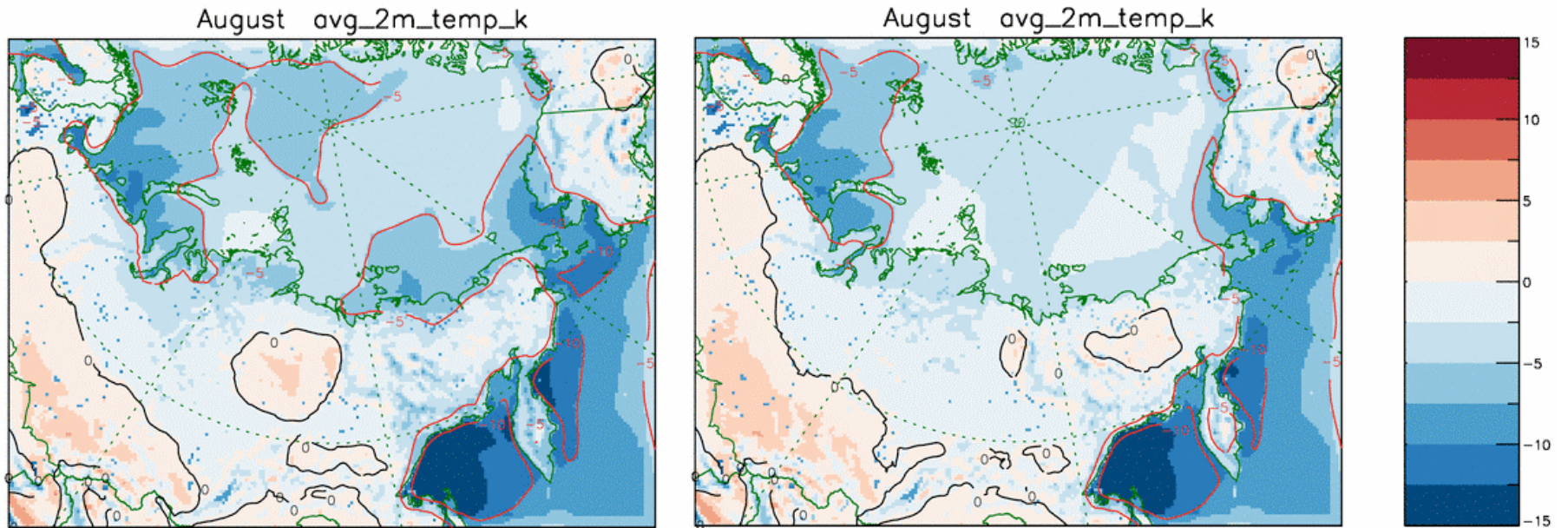
Thermal State – Boundary Conditions



OLD

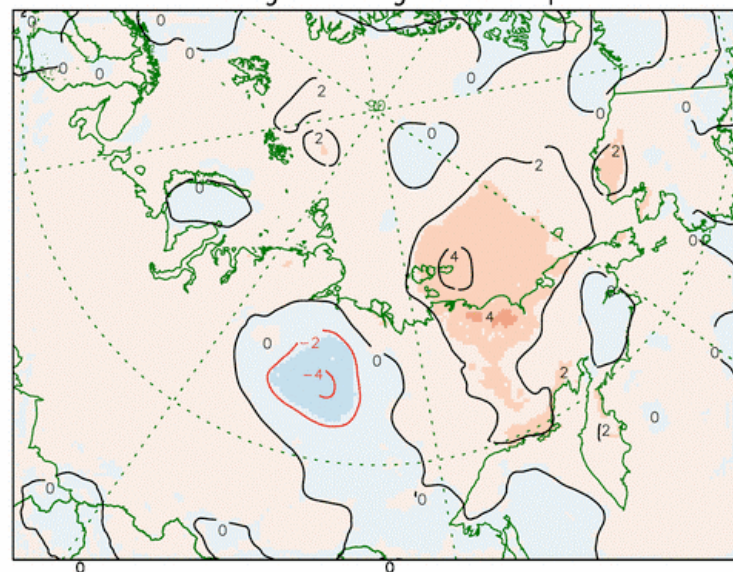
DIFFERENCE

WRF vs ERA-40 2-meter Temperature (AUGUST)



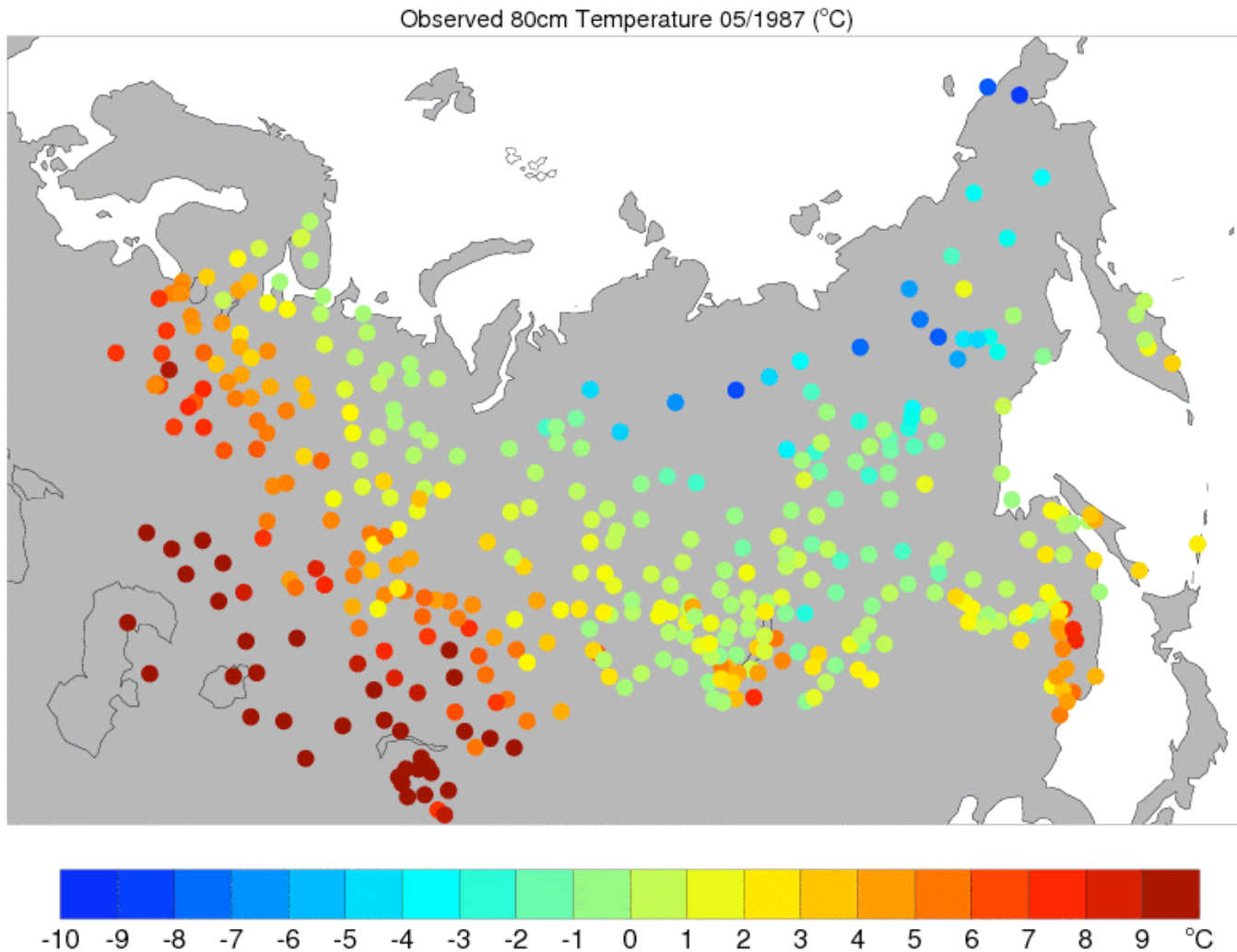
Control: WRF - ERA40

New LBC: WRF - ERA40



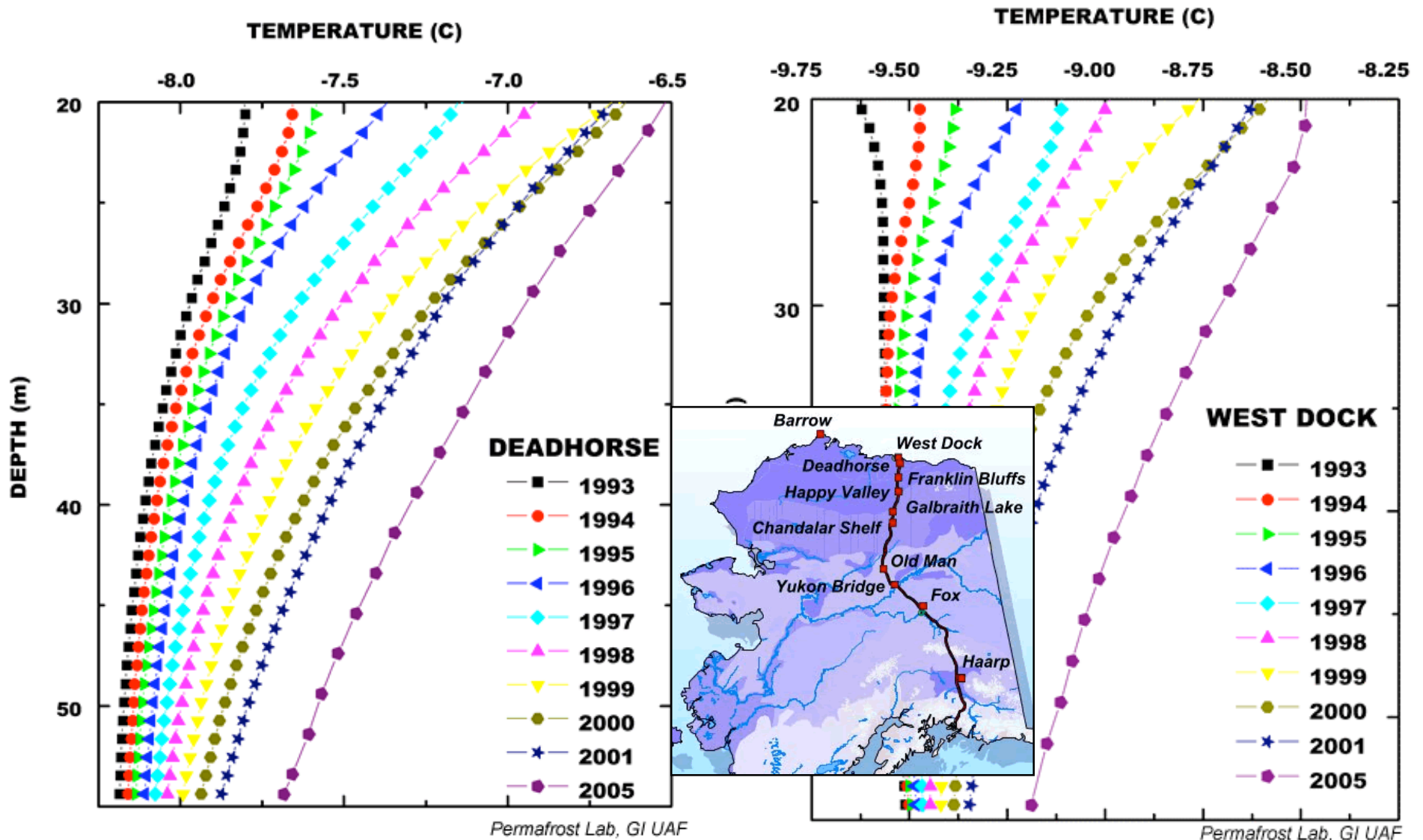
New - Control

Soil Station Data

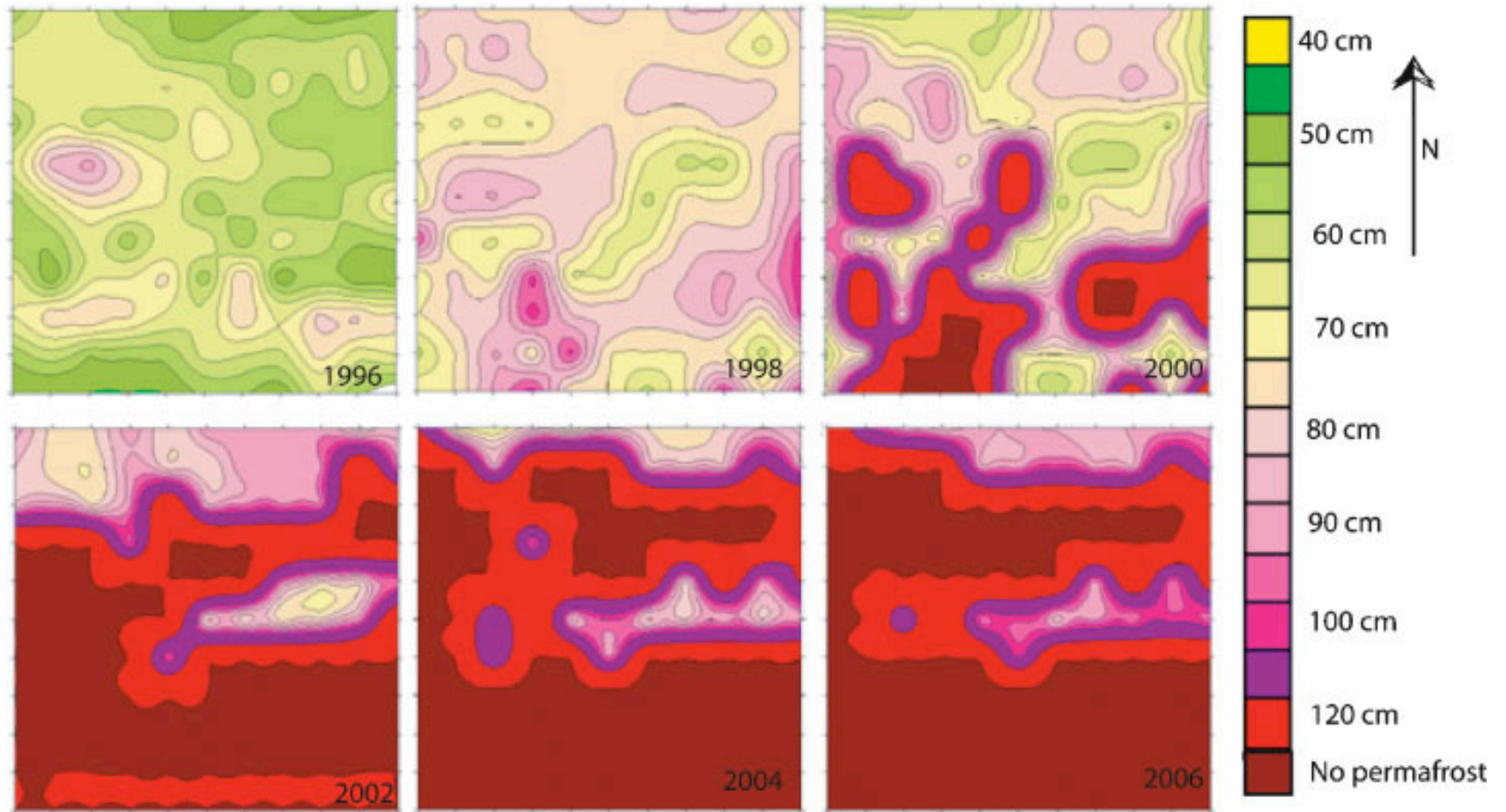


- Other regions?
- Land cover type?
- What level of accuracy is needed?
- Change over time

Recent Changes in Permafrost

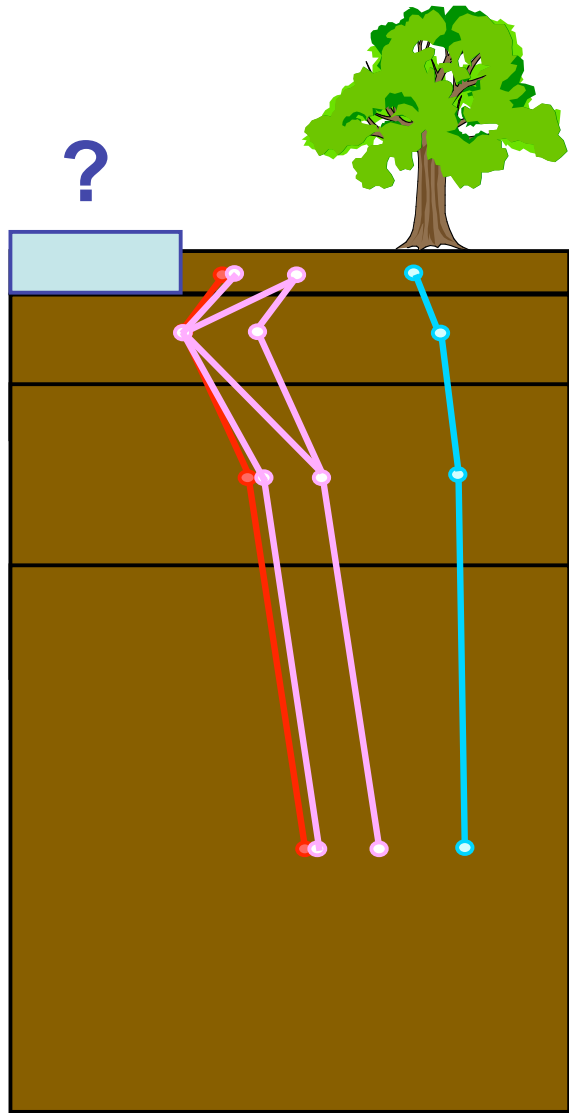


Recent Changes in Permafrost



Spatial and temporal variation in active-layer thickness at Katterjokk, Sweden

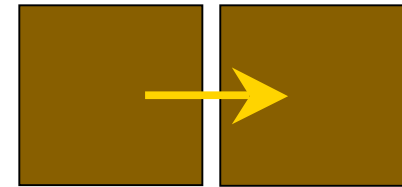
Information Propagation



X =

Observed T
Model T
Model Moisture

T₁
T₂
T₃
T₄
W₁
W₂
W₃
W₄



Information Propagation

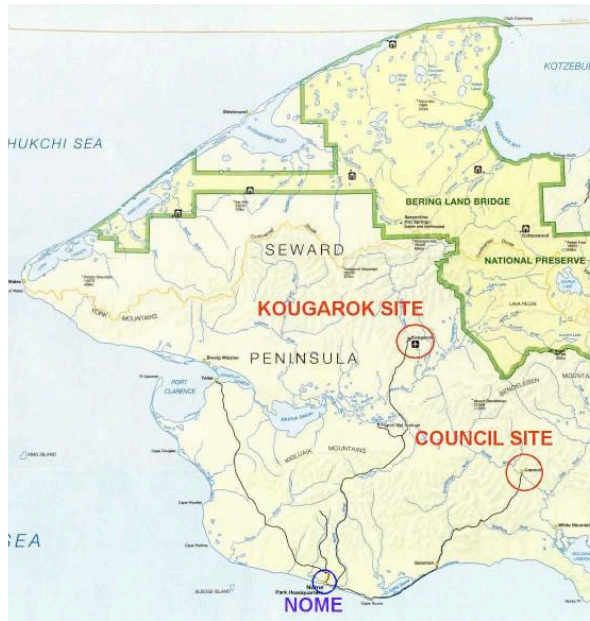
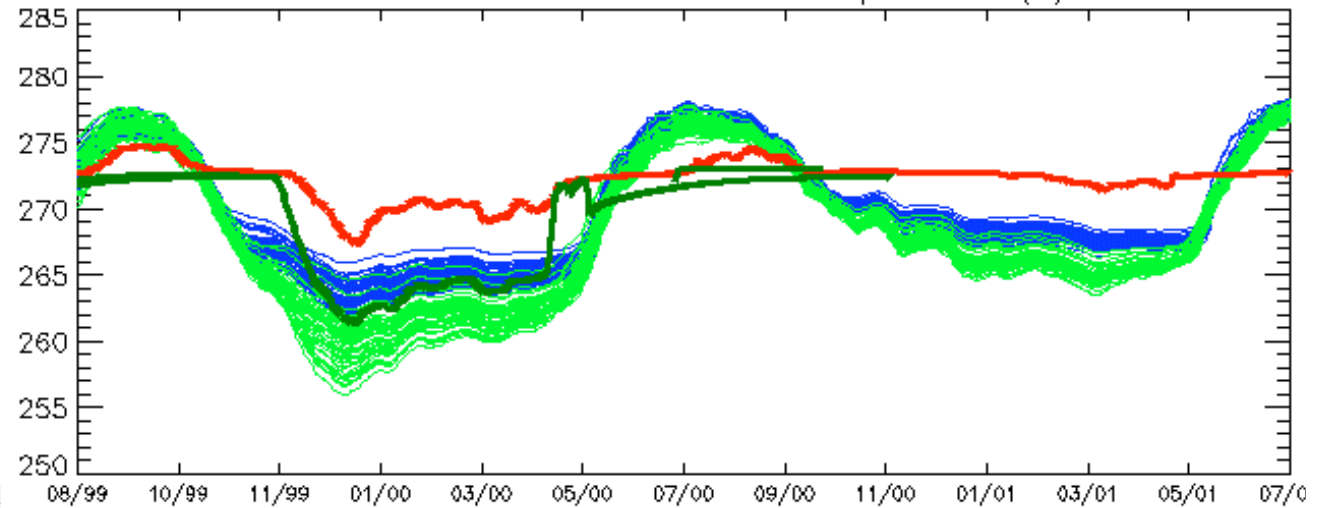
Council : Model

Council : Obs.

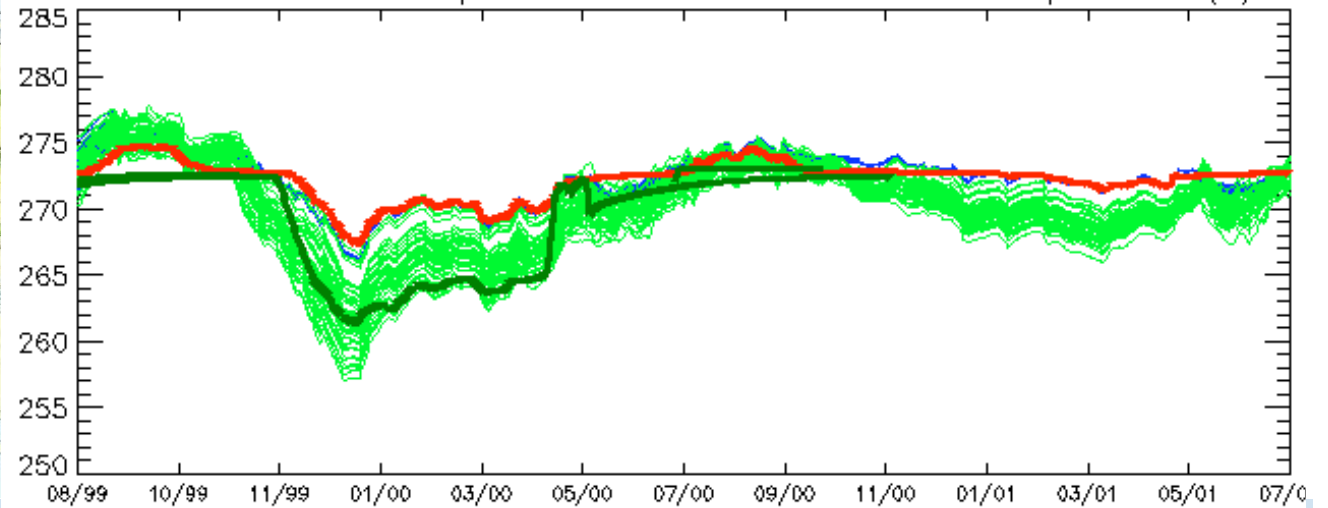
Kougarok : Model

Kougarok : Obs

Standard – 40–100cm Soil Temperature (K)



Assimilated Soil Temp @ 25cm – 40–100cm Soil Temperature (K)



Pan-Arctic Ground Thermal State Prospects

	Z (coverage)	H (Obs to Model)	R (error estimate)
Station Data (Temperature)	Spotty Day-Year	Point vs Area Soil type	Point vs Grid Box
Passive Microwave (Freeze/Thaw)	Global Daily?	Radiance Model	Snow, Veg

Future Data

- IPY improved monitoring (more stations)
- InSAR – active layer depth (experimental)
- LiDAR – expensive (experimental)
- SMAP – Level 4 product (2015, model + data assim)

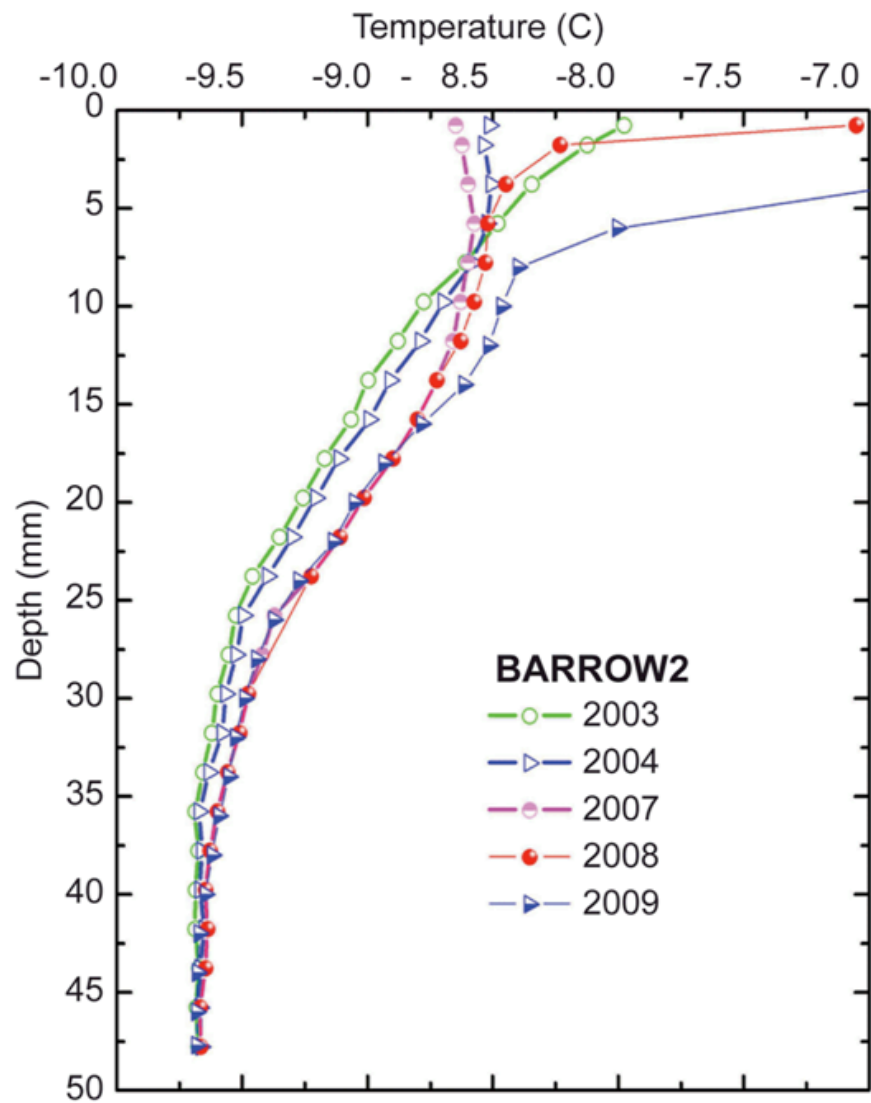
Land Data Assimilation Summary

- Methods being developed
- Data increasing, but less than ideal
- Heterogeneity a big problem

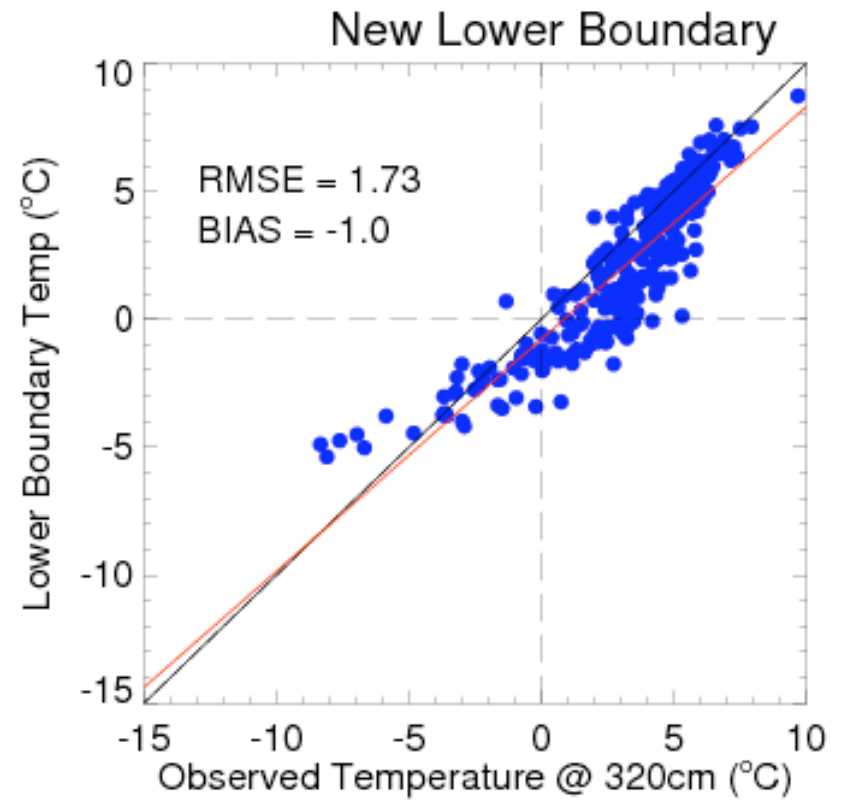
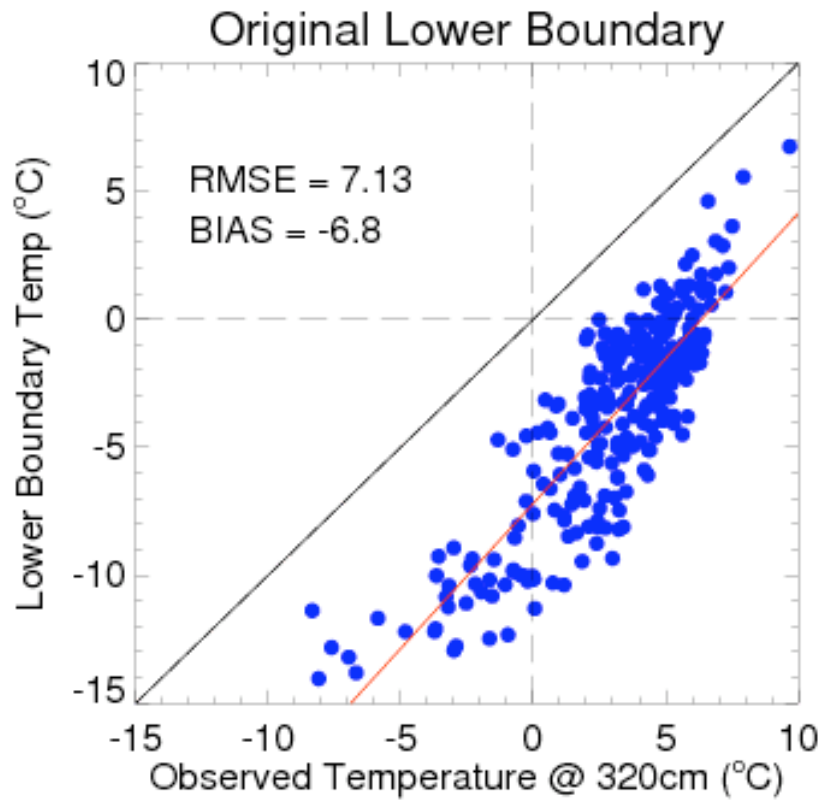
- Initialize model states (single variable, jointly)
- Innovation analysis for model deficiency
 - Parameter uncertainty
 - Structural problems

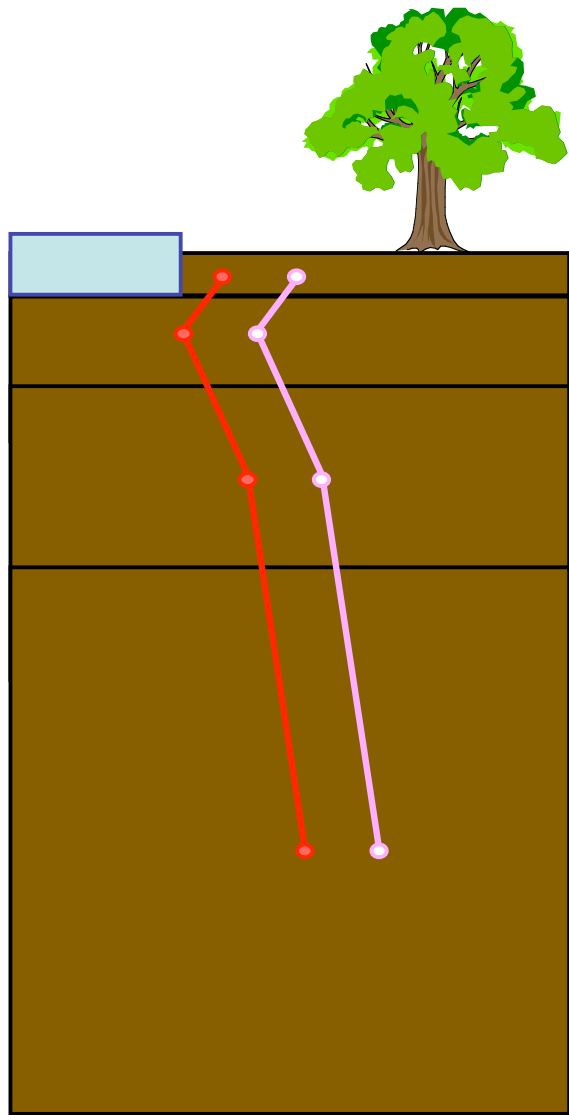
Thanks

Questions?



Validation in Eurasia





10cm

40cm

100cm

200cm