

**Arctic Sea-Ice Thickness: Evidence of Decline
from a Multiple Regression Analysis
Incorporating Long-Range Dependence**

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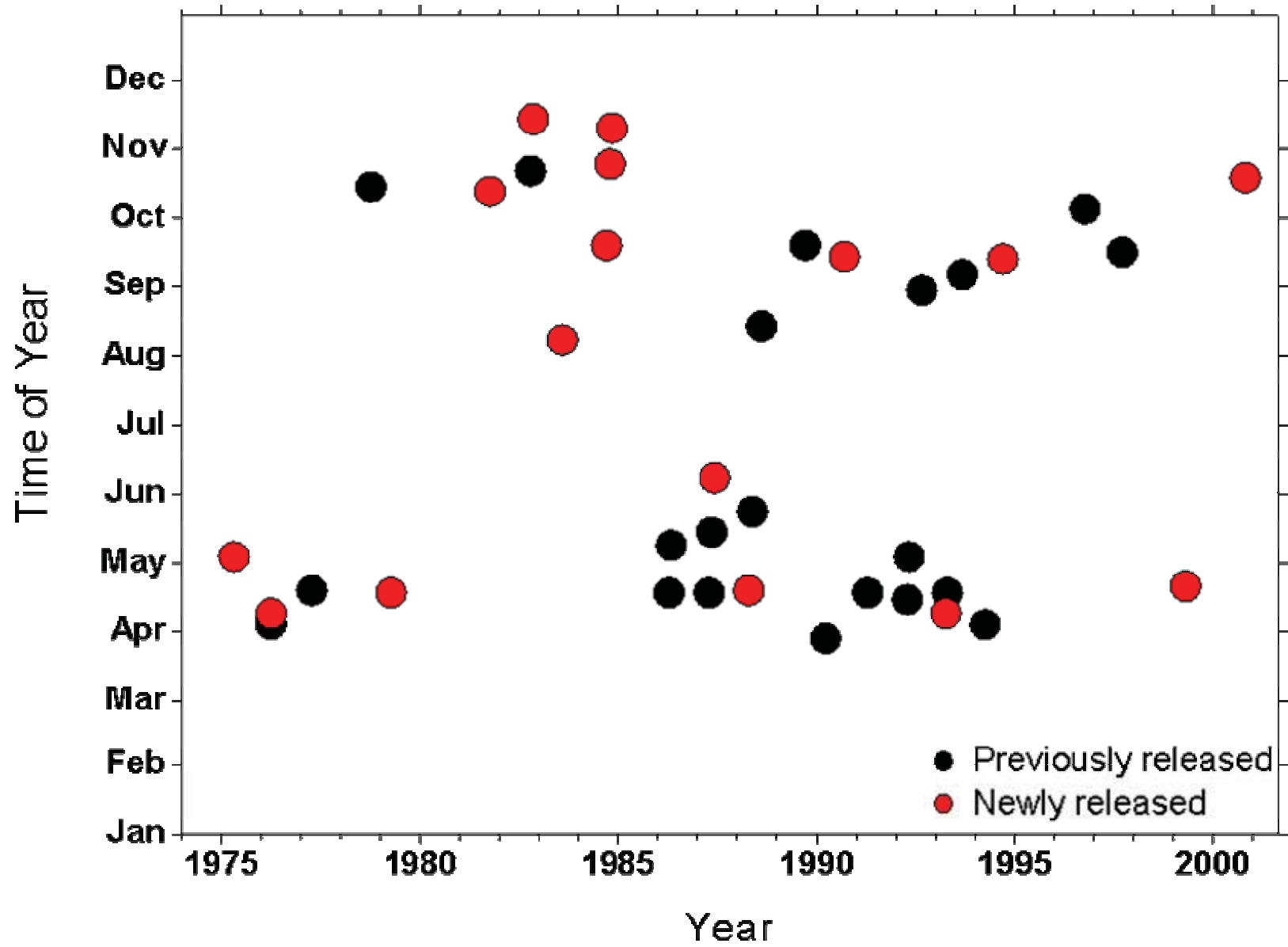
NSF-sponsored collaborative effort with Drew Rothrock (PI),
Tilmann Gneiting, Mark Wensnahan and Alan Thorndike

overheads for talk available at

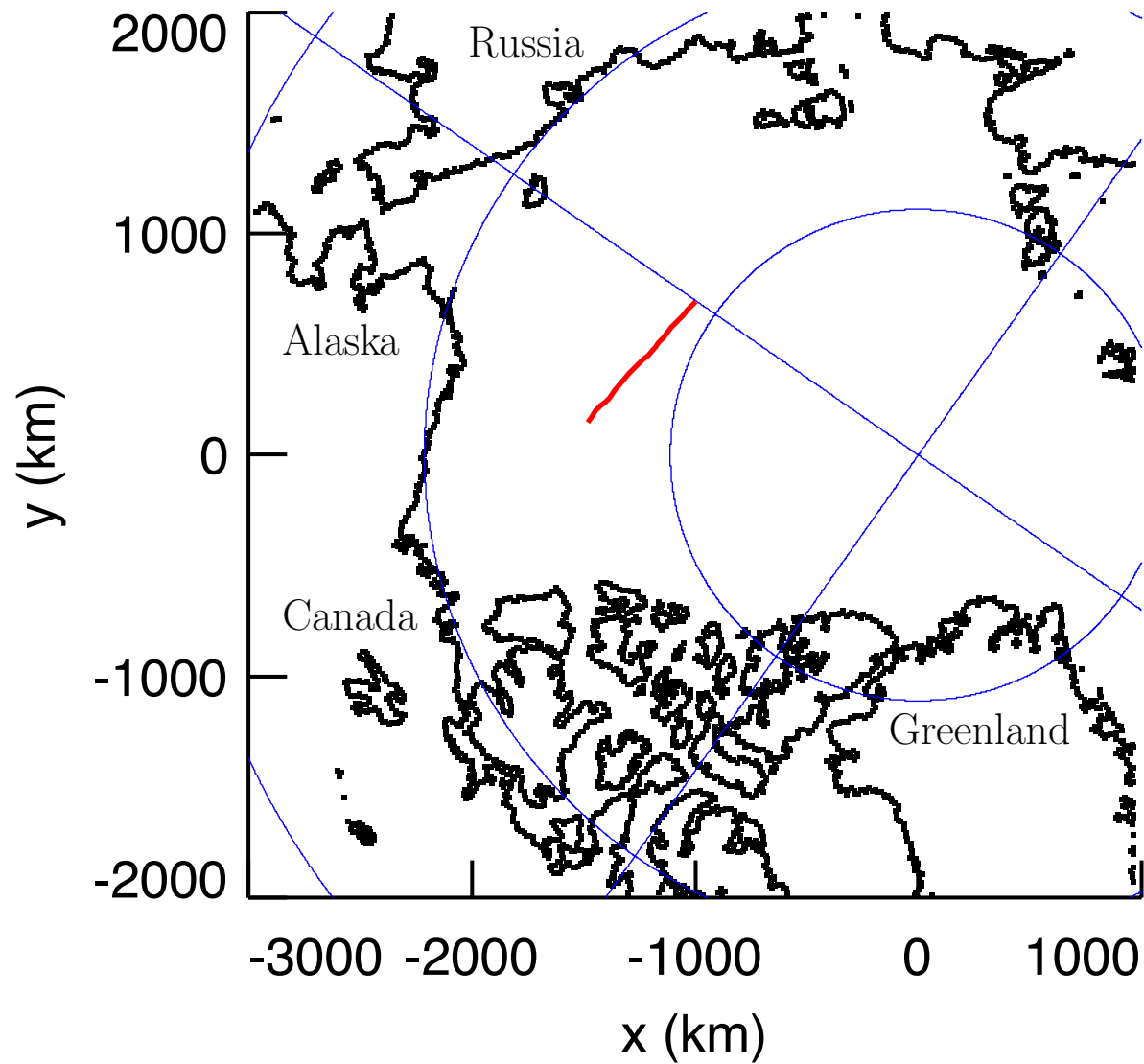
<http://faculty.washington.edu/dbp/talks.html>

Background

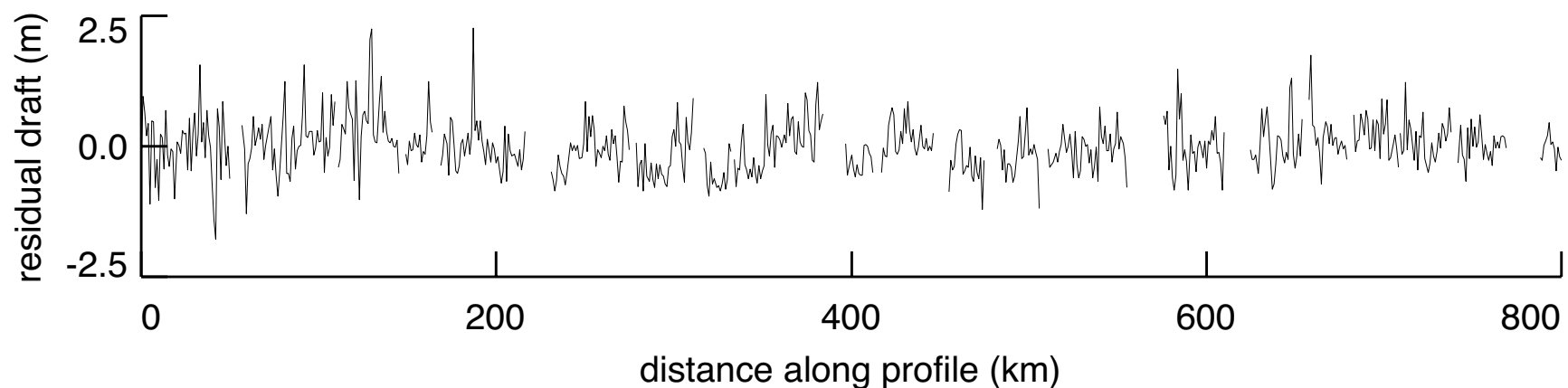
- scientific question of interest: has the average thickness of Arctic sea ice declined significantly over the past 30 years?
- thickness can be deduced from measurements of draft (submerged portion of sea ice)
- draft measured using upward-looking sonars on submarines
- our analysis differs from previous ones by use of
 - a new statistical model for draft measurements
 - newly archived data for submarine cruises from 1975 to 2001 (almost doubling the amount of available data)



Map of Arctic Region with **One Submarine Track**



Residual Draft Profile along One Submarine Track



- above comes from longest track from 1997 cruises
- draft measured every meter & used to form 1 km averages
- linear trend subtracted to form residual draft profile $\overline{H}_{1,n}$ (for profiles less than 200 km or so, need only subtract sample mean)
- residuals approximately Gaussian (room for improvement?)
- note: lots of gaps in draft profile (632 averages over 803 km)

Statistical Modeling of Draft Profiles

- simple model of independence for $\overline{H}_{1,n}$ along profile not viable (e.g., adjacent measurements $\overline{H}_{1,n}$ and $\overline{H}_{1,n+1}$ are correlated)
- assume $\overline{H}_{1,n}$ is a realization of a zero mean Gaussian stationary process (a ‘time’ series with distance replacing ‘time’)
- process fully characterized by its variance σ_1^2 and autocorrelation sequence $\rho_d \equiv E\{\overline{H}_{1,n}\overline{H}_{1,n+d}\}/\sigma_1^2$
- consider two simple parametric forms for ρ_d corresponding to
 - first-order autoregressive (AR(1)) process
 - fractionally differenced (FD) process

First-Order Autoregressive (AR(1)) Processes

- process satisfies

$$\bar{H}_{1,n} = \phi \bar{H}_{1,n-1} + \epsilon_n = \sum_{j=0}^{\infty} \psi_j \epsilon_{n-j} \text{ with } \psi_j = \phi^j,$$

where $|\phi| < 1$, and ϵ_n is Gaussian white noise with variance σ_ϵ^2

- have $\rho_d = \phi^{|d|}$
- related to a first-order stochastic differential equation with ‘correlation time’ dictated by ϕ
- widely used in climate research to model time series
- given gappy draft profiles, can estimate ϕ and σ_ϵ^2 using maximum likelihood (Jones, 1980), yielding $\hat{\phi} \doteq 0.36 (\pm 0.04)$

Fractionally Differences (FD) Processes

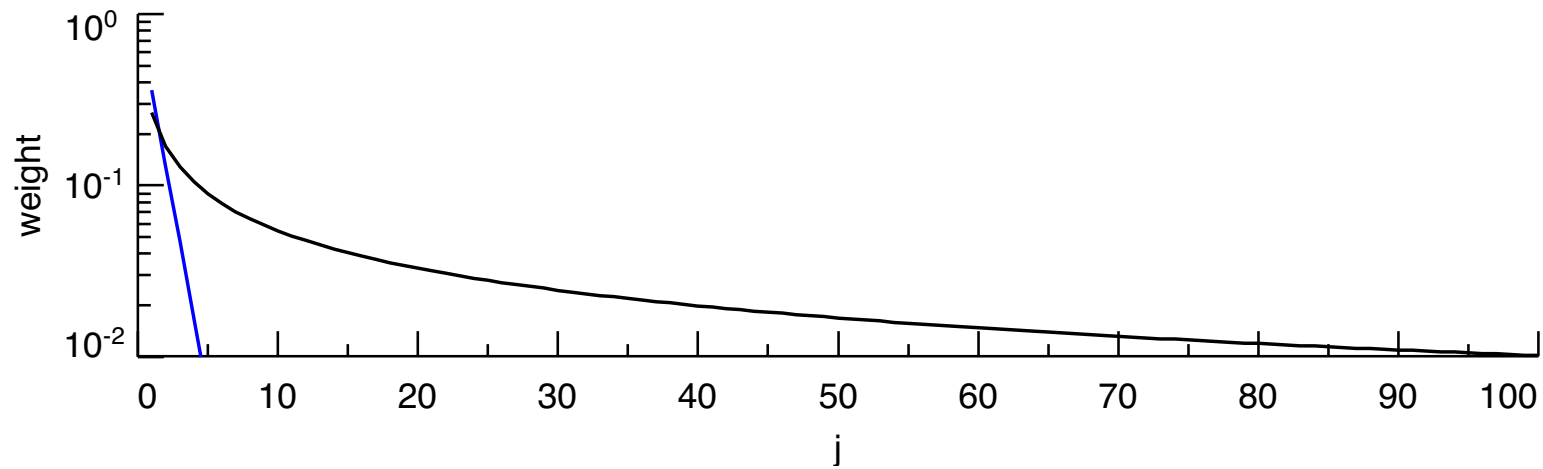
- process satisfies

$$\overline{H}_{1,n} = \sum_{j=0}^{\infty} \psi_j \epsilon_{n-j} \quad \text{with} \quad \psi_j = \frac{\Gamma(j + \delta)}{\Gamma(j + 1)\Gamma(\delta)},$$

where $|\delta| < 1/2$, and ϵ_n is as before

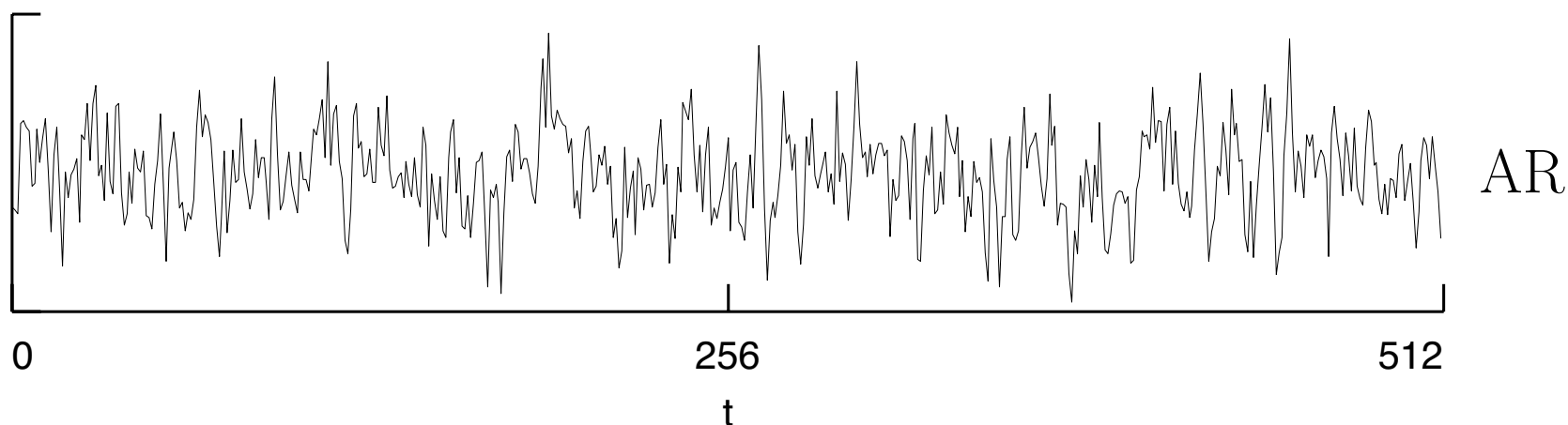
- have $\rho_d \approx C|d|^{2\delta-1}$ (much slower decay rate than for AR(1))
- related to average of many first-order stochastic differential equations with different correlation times
- popular model for ‘long-range’ (or ‘long-memory’) dependence
- given gappy profiles, can estimate δ and σ_ϵ^2 using maximum likelihood (Palma & Chan, 1997), yielding $\hat{\delta} \doteq 0.27 (\pm 0.03)$
- Q: how do these two models compare?

Qualitative Comparison I: ψ Weights



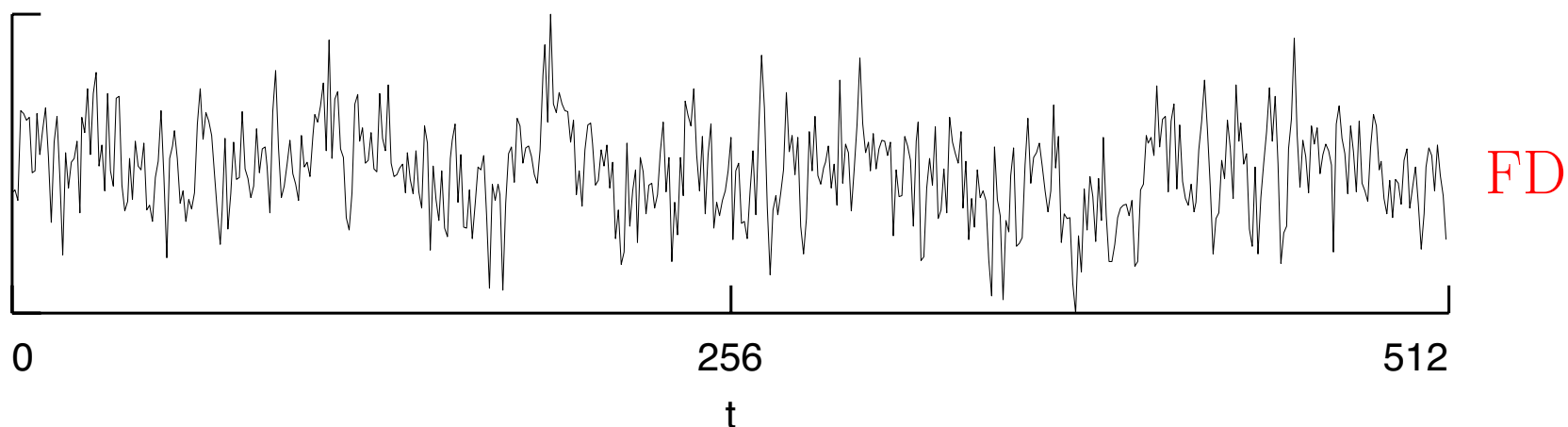
- weights ψ_j used to create AR with $\phi = 0.36$ (blue curve) and FD with $\delta = 0.27$ (black curve) processes from a weighted average of white noise

Qualitative Comparison II: Simulated Draft Profiles



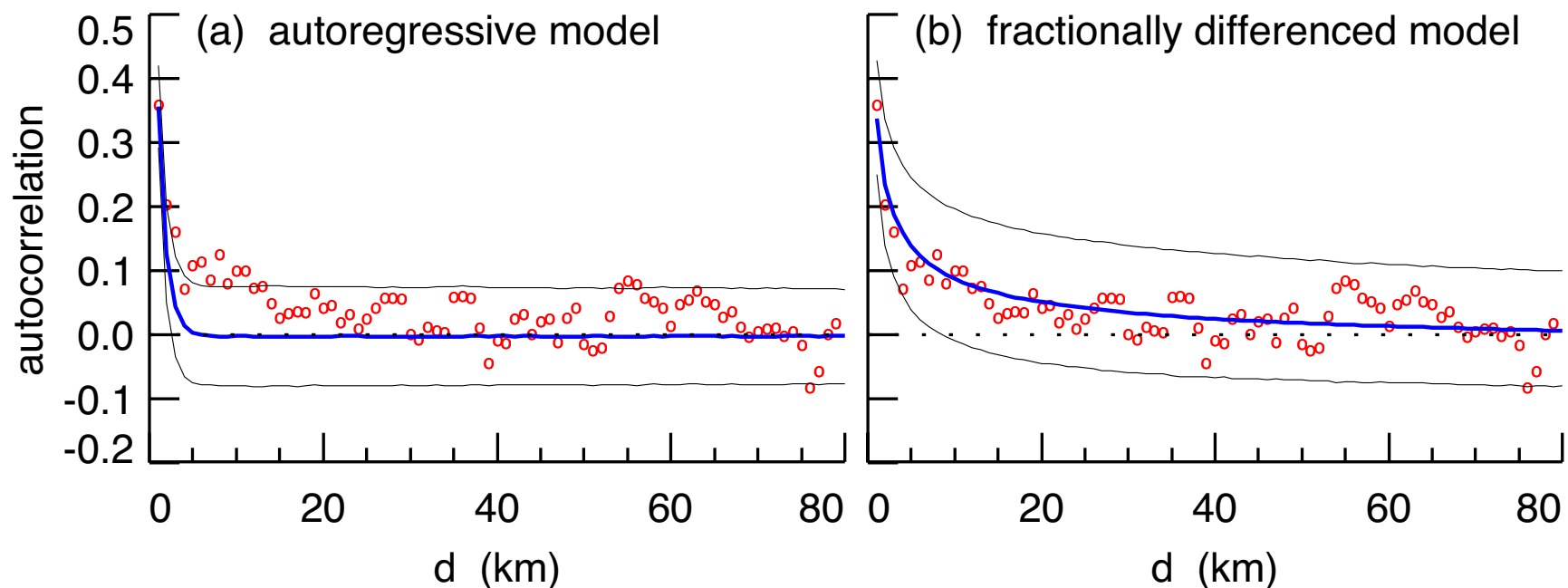
- consider simulated AR ($\phi = 0.36$) & **FD** ($\delta = 0.27$) profiles
- simulations formed using circulant embedding technique that maps same 1024 IID Gaussian deviates to both profiles

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Third Comparison: Autocorrelation Sequences



- sample (circles) and theoretical sequences (middle curves)
- upper and lower curves are 95% pointwise confidence intervals for ρ_d assuming relevant model (AR or FD)

Fourth Comparison: Variance of Sample Means

- given $\bar{H}_{1,n}$, $n = 0, \dots, N - 1$, consider statistical properties of length L averages

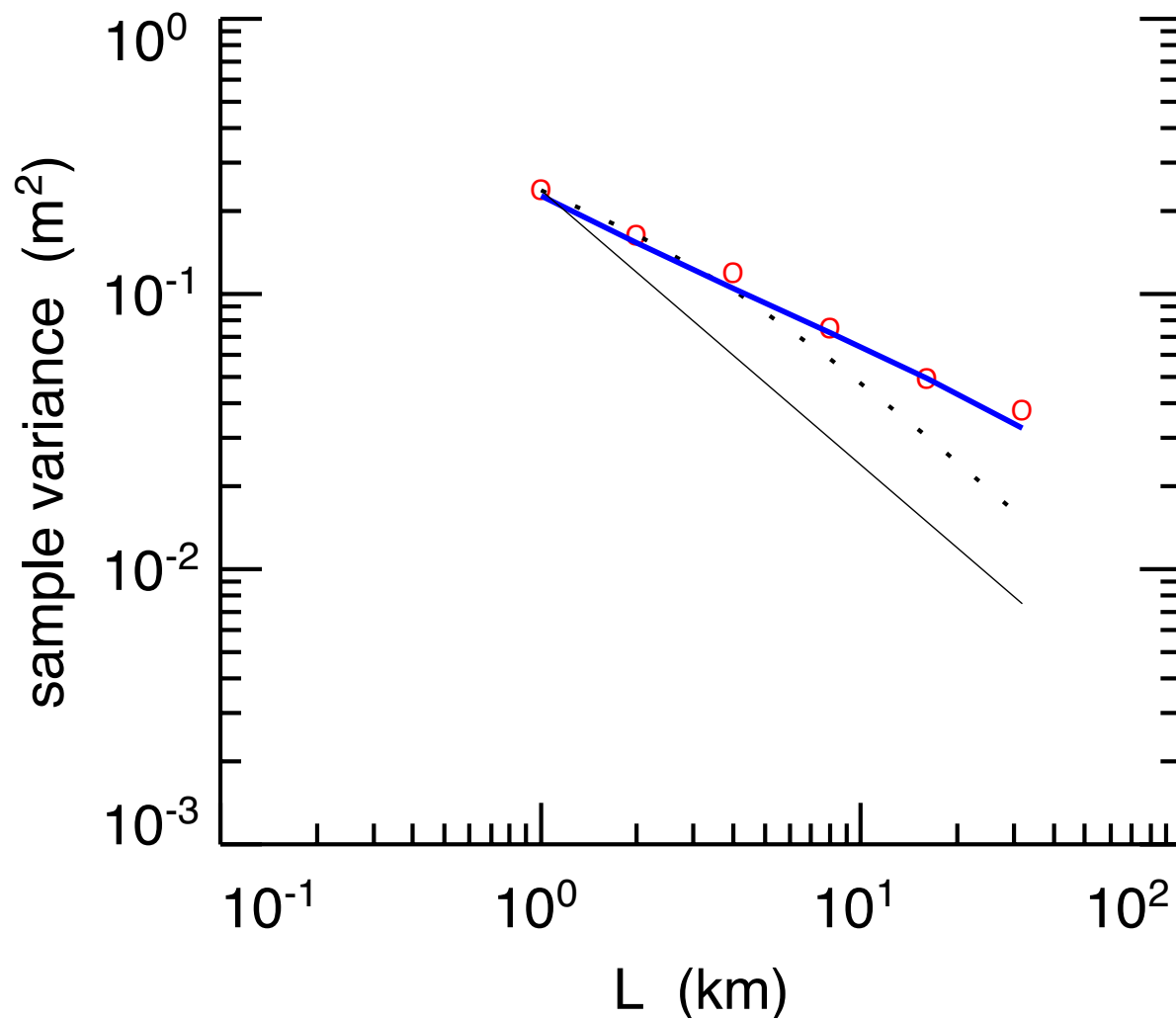
$$\bar{H}_{L,m} = \frac{1}{L} \sum_{l=0}^{L-1} \bar{H}_{1,mL+l}, \quad m = 0, 1, \dots, \lfloor N/L \rfloor - 1$$

- let $\sigma_L^2 = \text{var} \{ \bar{H}_{L,m} \}$
- for AR and FD models, have

$$\sigma_L^2 \approx \sigma_1^2 \times \frac{1 + \phi}{1 - \phi} \times L^{-1} \quad \text{and} \quad \sigma_L^2 \approx \sigma_1^2 \times \frac{\Gamma(1 - \delta)}{(2\delta + 1)\Gamma(1 + \delta)} \times L^{-1+2\delta}$$

- can compare sample estimates $\hat{\sigma}_L^2$ with $E\{\hat{\sigma}_L^2\}$ for various L

Sample $\hat{\sigma}_L^2$ (Circles) and Theoretical σ_L^2 versus L

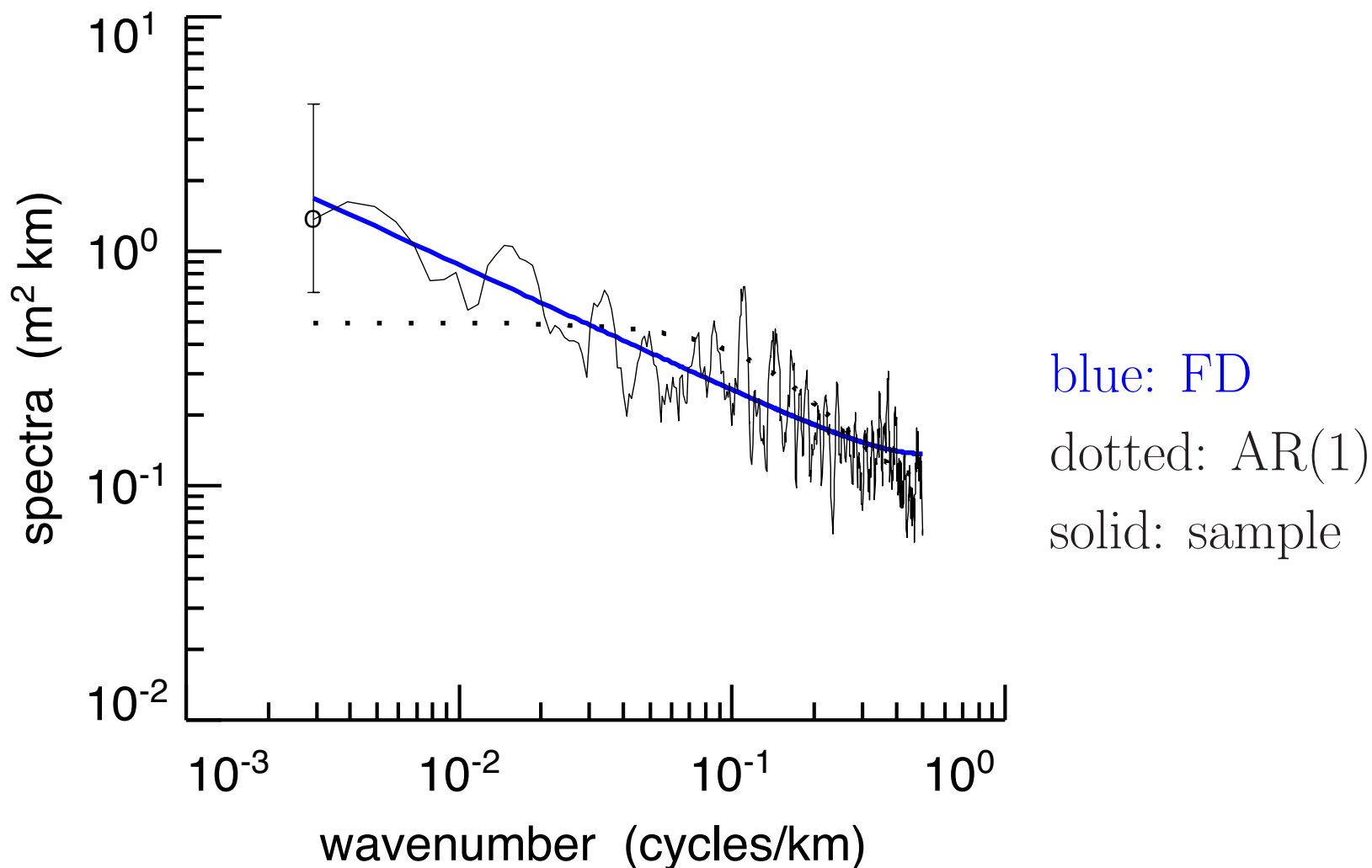


blue: FD

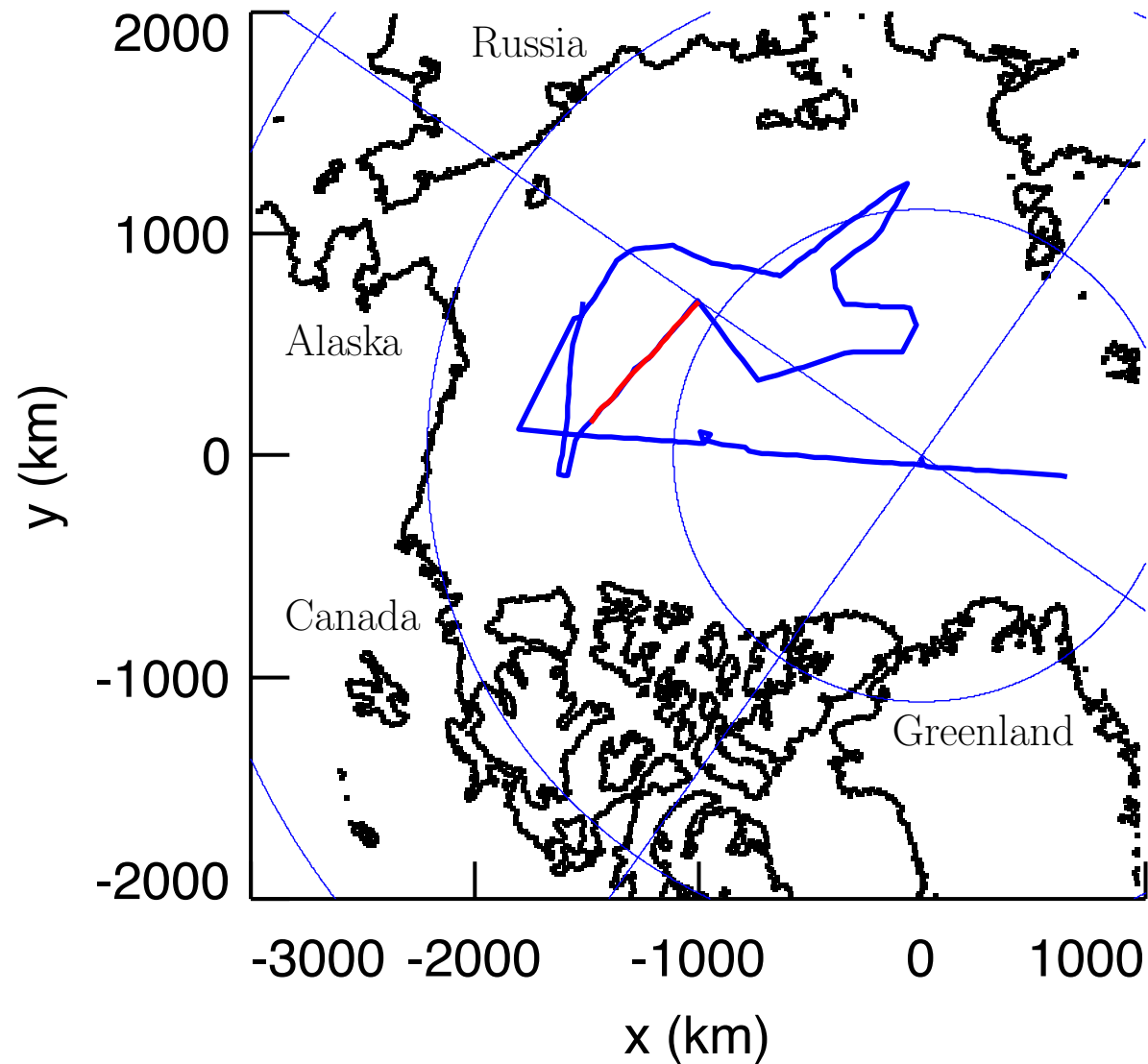
dotted: AR(1)

solid: white noise

Fifth Comparison: Sample and Theoretical Spectra



Map of Arctic Region with Tracks Taken in 1997



Spatial Model for One Kilometer Averages

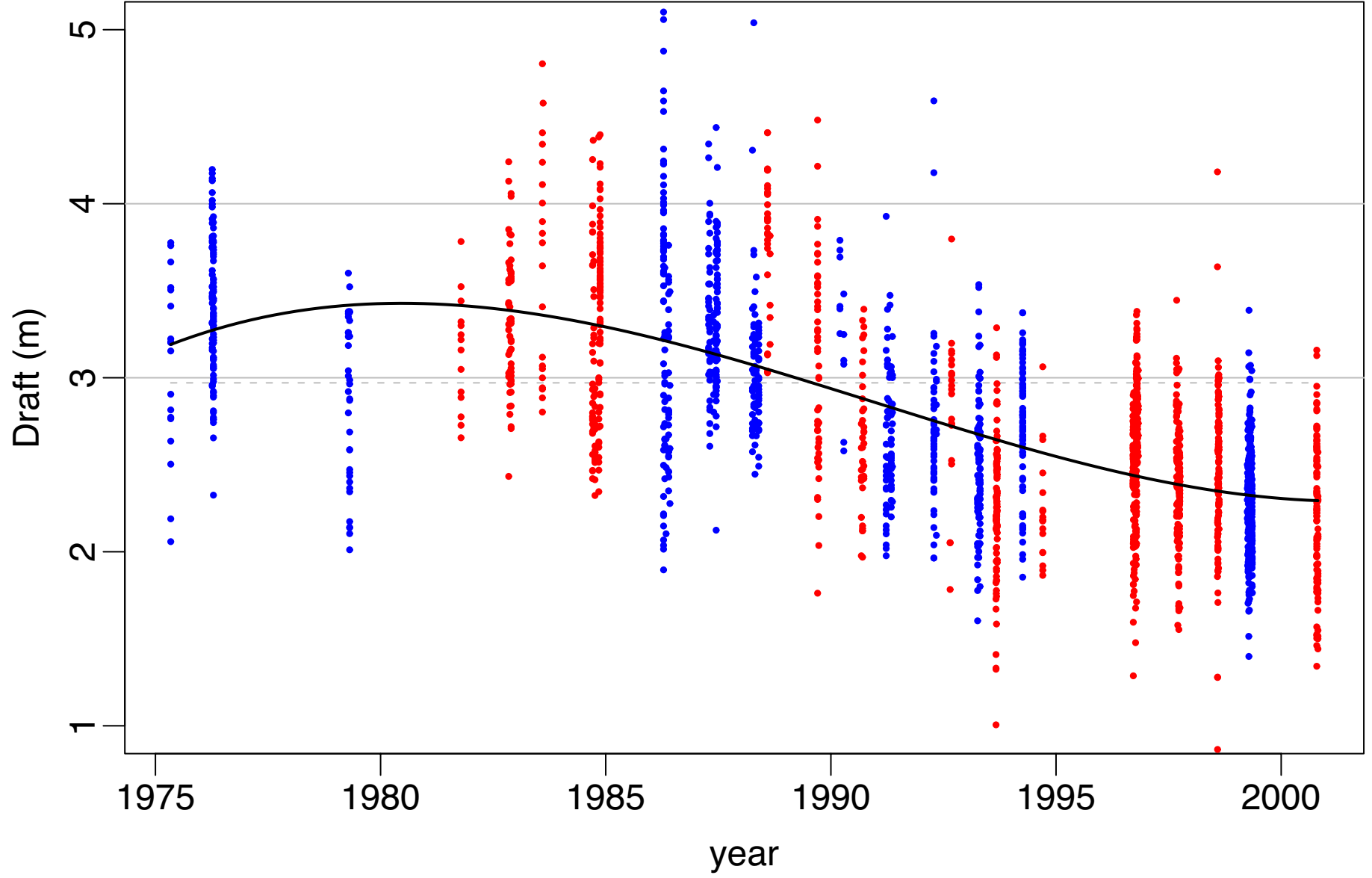
- analysis of additional profiles in 1997 and other years indicates FD model with $\delta = 0.27$ is generally viable
- can regard $\overline{H}_{1,n}$ as samples from a stationary and isotropic two-dimensional (2D) random process with covariances given by

$$\text{cov} \{ \overline{H}_{1,\mathbf{x}_n}, \overline{H}_{1,\mathbf{x}_n+\mathbf{d}} \} \equiv \sigma_1^2 \times \frac{\Gamma(|\mathbf{d}| + \delta)\Gamma(1 - \delta)}{\Gamma(|\mathbf{d}| + 1 - \delta)\Gamma(\delta)},$$

where \mathbf{x}_n is a 2D vector indicating the location of the 1 km average, and \mathbf{d} is an arbitrary 2D vector

Multiple Regression Model

- let $\overline{H}_{1,\mathbf{x}_n,t}$ represent average of 1 km measurements taken at location \mathbf{x}_n and time t ($\mathbf{x}_n = [0, 0] = \text{Pole}$ & $t \in [1975, 2001]$)
- let τ represent the time of year (i.e., $\tau = t \bmod 1$)
- assume model $\overline{H}_{1,\mathbf{x}_n,t} = C + I(t) + A(\tau) + S(\mathbf{x}_n) + \epsilon_{\mathbf{x}_n,t}$, where
 - C is a constant
 - $I(t)$ is the interannual variation (cubic polynomial)
 - $A(\tau)$ is the annual cycle (cosine, unknown amplitude & phase)
 - $S(\mathbf{x}_n)$ is the spatial field (fifth order polynomial)
 - $\epsilon_{\mathbf{x}_n,t}$ is an error term dictated by FD model within a given season (different seasons/years assumed independent)
- used generalized least squares to fit model, with order of polynomials dictated by t -tests



Interannual Variation $I(t)$

- residuals shown about fitted $I(t)$ (blue for January to June data, red for rest of year)
- change from 1981 to 2000 is -1.13 m
- steepest decline (-0.08 m/yr) occurred in 1991
- no recovery by 2000
- much fuller data set strengthens previous results (Rothrock *et al.*, 1999, and Tucker *et al.*, 2001)

Concluding Remarks

- multiple regression model explains 79% of variance in data (standard deviation is 0.98 m)
- unexplained variance has standard deviation of 0.46 m
- estimated standard deviation of measurement errors is 0.25 m
- improvements ('polishing the cannon ball'):
 - relax assumption of a constant spatial field across time
 - estimate δ from spatial data, not from profiles

Main References (1–3) & Additional Ones (4–7)

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7. W. B. Tucker, J. W. Weatherly, D. T. Eppler, L. D. Farmer and D. L. Bentley (2001), ‘Evidence for Rapid Thinning of Sea Ice in the Western Arctic Ocean at the End of the 1980s,’ *Geophysical Research Letters*, **28**, pp. 2851–2854.