

Efficient kriging for real-time spatio-temporal interpolation

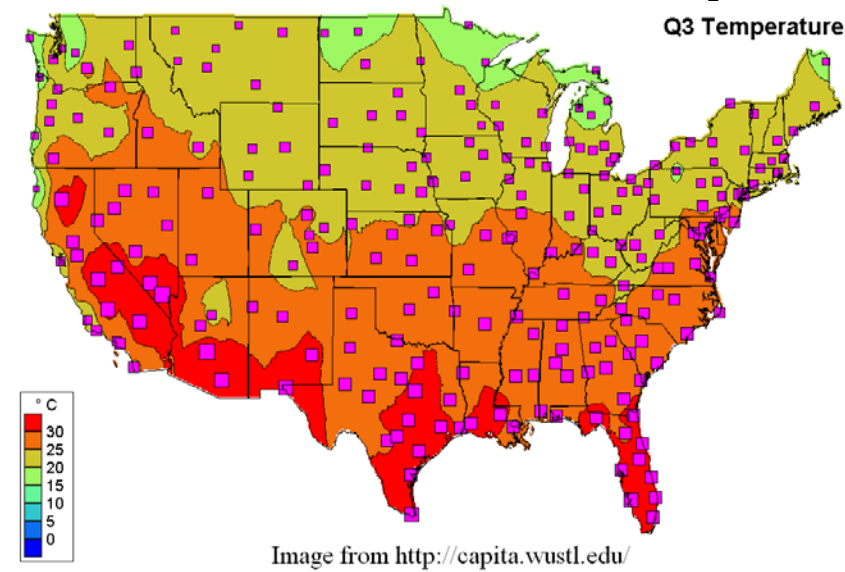
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Kriging

- In atmospheric science, data is often collected at scattered station locations.
- Analysis require the data to be regularly-gridded or available at any query location of interest
- Need a scheme to interpolate scattered station data.



- Kriging** (Isaaks et al.) is a group of geostatistical techniques to interpolate the value of a random field (e.g., the temperature as a function of the geographic location) at an unobserved location from observations of its value at nearby locations.
- Types: simple, ordinary, universal, indicator, disjunctive, IRFk, lognormal, etc.
- Ordinary kriging → Best Linear Unbiased Estimator

Ordinary kriging - formulation

- Given: Observed recordings $\{v_1, v_2, \dots, v_N\}$ at locations $\{x_1, x_2, \dots, x_N\}$
- Estimate $\{v_1^*, v_2^*, \dots, v_M^*\}$ at $\{x_1^*, x_2^*, \dots, x_M^*\}$

- Linear: $v_j^* = \sum w_{ij} v_i$
- Unbiased: $E(v_j^*) = E(v_j) \rightarrow \sum_i w_{ij} = 1$, for all j
- Best → least variance

$$C_{i,k*} = \sum_{j=1}^N w_j C_{ij} + \mu$$

- C_{ij} = covariance between x_i and x_j
- Given by functional representation
 - Commonly used: Gaussian
 - $C_{ij} = \exp(-\|x_i - x_j\|^2 / h^2) + \delta(i, j)$
- $\mu \rightarrow$ Lagrange multiplier

Ordinary kriging - system

- For each $k^* = 1, 2, \dots, M$, solve

$$\begin{pmatrix} C_{11} & \dots & C_{1n} & 1 \\ \vdots & \ddots & \vdots & \vdots \\ C_{n1} & \dots & C_{nn} & 1 \\ 1 & \dots & 1 & 0 \end{pmatrix} \times \begin{pmatrix} w_1 \\ \vdots \\ w_N \\ \mu \end{pmatrix} = \begin{pmatrix} C_{1k*} \\ \vdots \\ C_{Nk*} \\ 1 \end{pmatrix}$$

Kriging computational cost

- $O(N^4)$ complexity

$$\begin{pmatrix} v_{1*} \\ \vdots \\ v_{M*} \end{pmatrix} = \begin{pmatrix} v_1 \\ \vdots \\ v_N \\ 0 \end{pmatrix}^T \begin{pmatrix} C_{11} & \dots & C_{1N} & 1 \\ \vdots & \ddots & \vdots & \vdots \\ C_{N1} & \dots & C_{NN} & 1 \\ 1 & \dots & 1 & 0 \end{pmatrix}^{-1} \begin{pmatrix} C_{1*1} & \dots & C_{1*N} \\ \vdots & \ddots & \vdots \\ C_{M*1} & \dots & C_{M*N} \\ 1 & \dots & 1 \end{pmatrix}$$

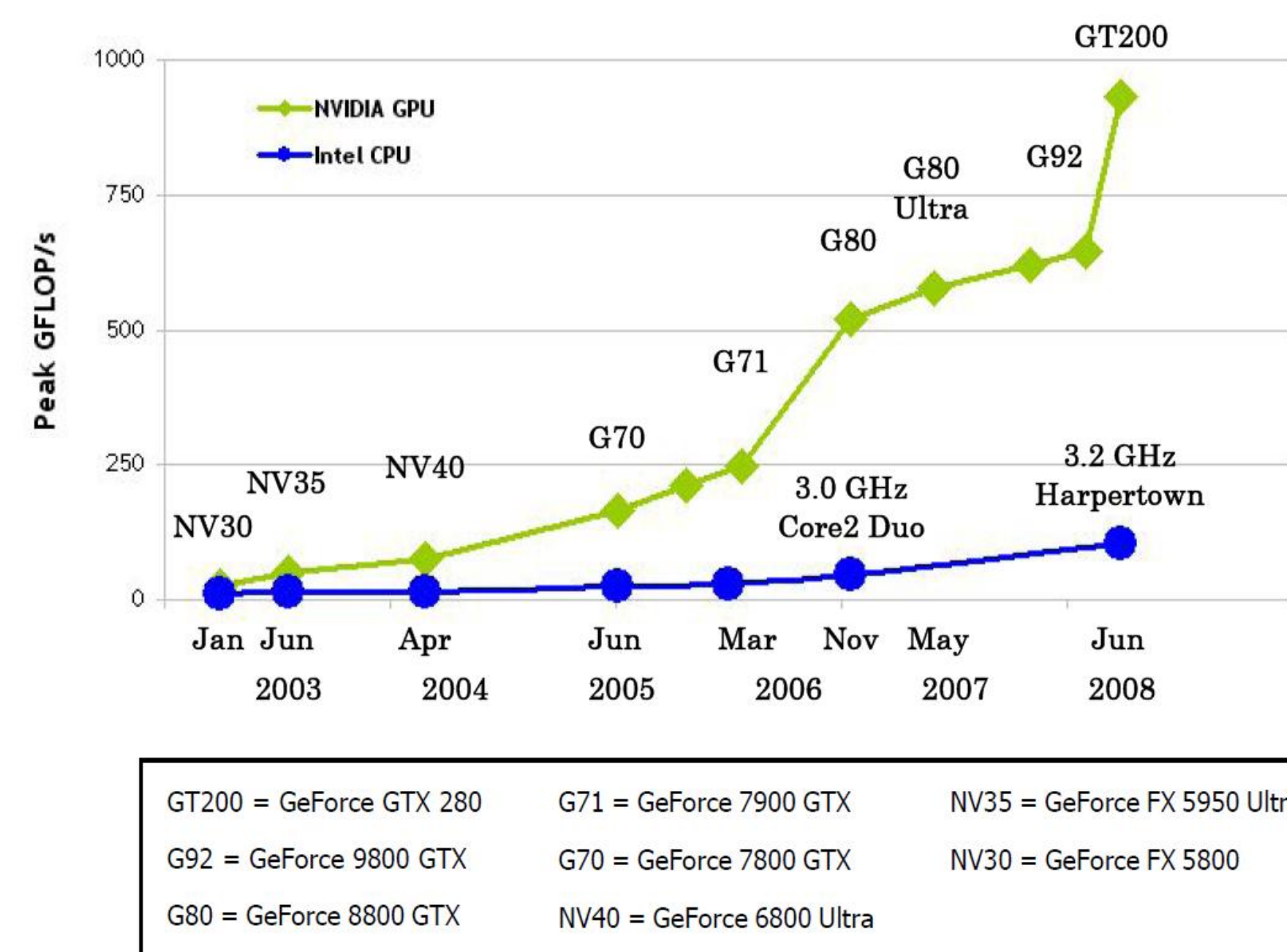
- $O(N^3)$ complexity

$$\begin{pmatrix} v_{1*} \\ \vdots \\ v_{M*} \end{pmatrix} = \begin{pmatrix} C_{1*1} & \dots & C_{1*N} & 1 \\ \vdots & \ddots & \vdots & \vdots \\ C_{M*1} & \dots & C_{M*N} & 1 \end{pmatrix} \begin{pmatrix} C_{11} & \dots & C_{1N} & 1 \\ \vdots & \ddots & \vdots & \vdots \\ C_{N1} & \dots & C_{NN} & 1 \\ 1 & \dots & 1 & 0 \end{pmatrix}^{-1} \begin{pmatrix} v_1 \\ \vdots \\ v_N \\ 0 \end{pmatrix}$$

- Iterative solvers → $O(kN^2)$
 - Saddle-point problem
 - Used SymmLQ (Paige et al.)

Graphical processing units (GPU)

- Graphics processors were developed to cater to the demands of real-time high definition graphics.
- Graphics processing units (GPU) are highly parallel, multi-core processor with tremendous computational horsepower and high memory bandwidth.

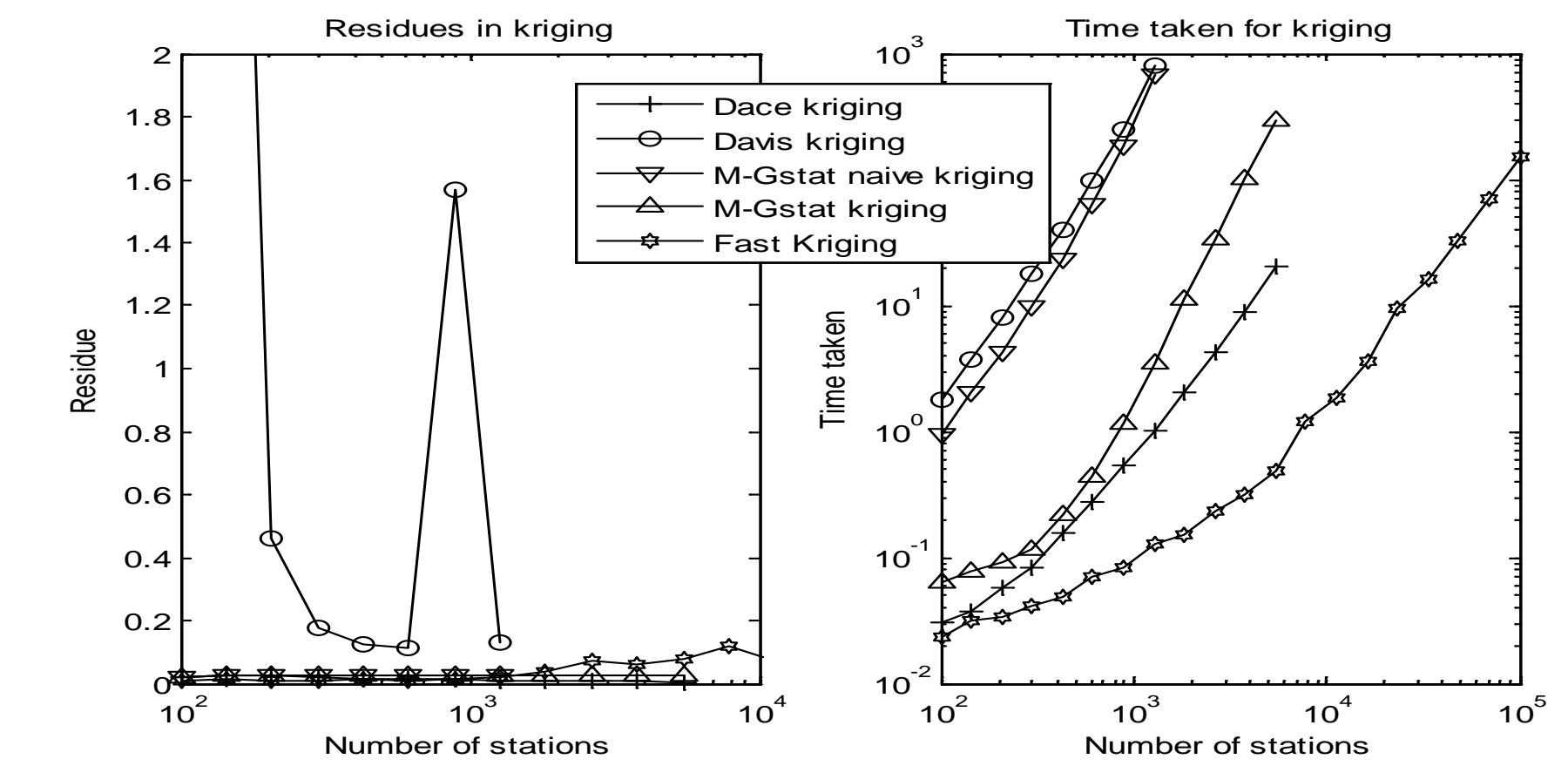


GPU-accelerated kriging

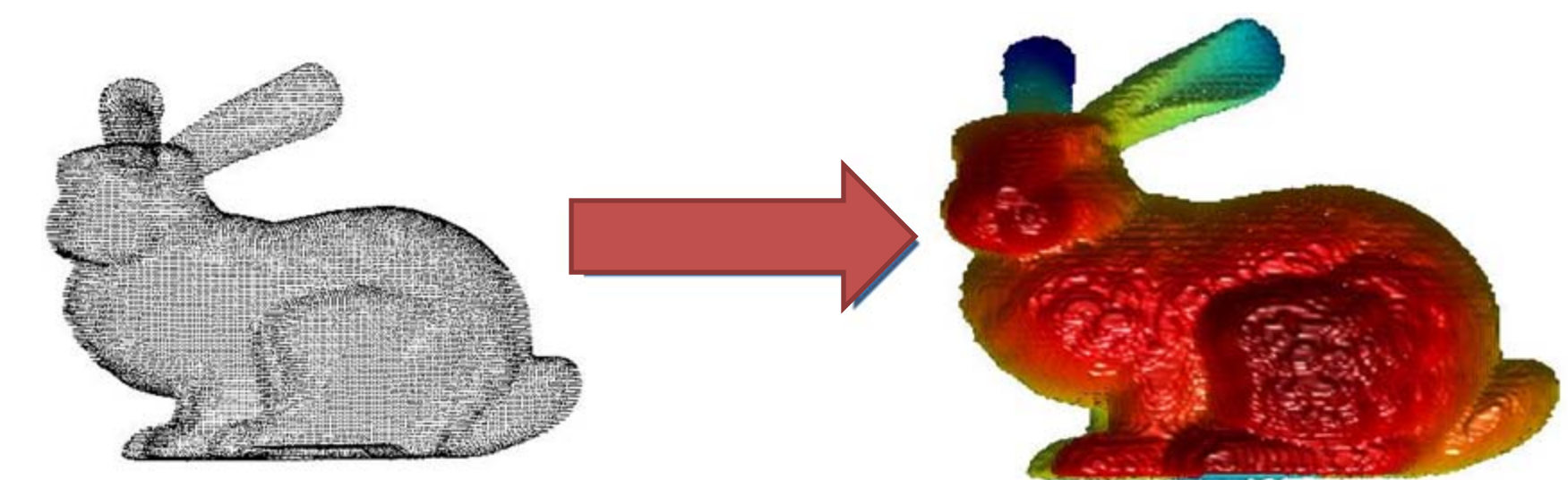
- Used the open source GPUML (Srinivasan et al.)
- Speedup: ~400X over a direct implementation.

Performance comparison

- Fastest among many available kriging and fitting tools



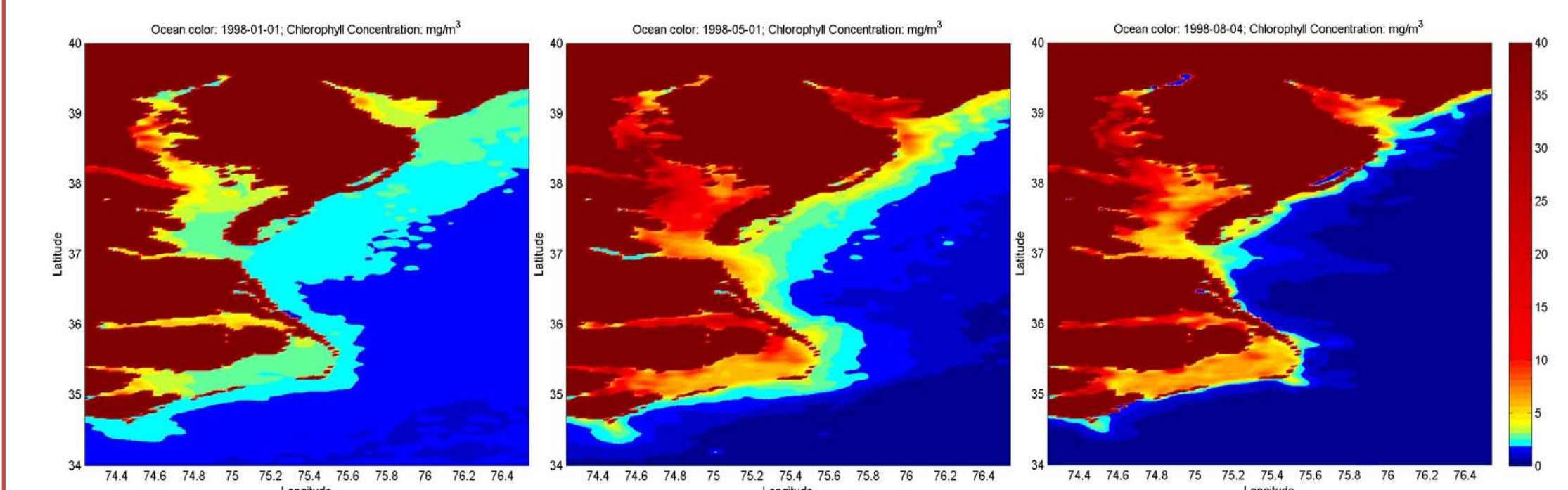
Spatial example



- Kriged at 8 million points using 100,000 points in 7 minutes, a direct implementation would take almost 2 days

Chesapeake bay ocean color

- Spatio-temporal kriging
- Kriged using 288 x 240 7-day gridded data
- ~ 70-80s to krig for 1 day prediction (direct implementation would take approximately 8 hours)



References

- Isaaks, E and Srivastava R, "Applied Geostatistics", Oxford University Press, New York, 542 pp, 1989.
- Paige, CC and Saunders, MA, "Solution of sparse indefinite systems of linear equations", SINUM 12, 617--629, 1975.
- Srinivasan BV and Duraiswami R, "Scaling kernel machine learning algorithm via the use of GPUs", GPU Technology Conference, NVIDIA Research Summit, 2009.