

How Long Can an Atmospheric Model Predict?



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References

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Physical Reality

- \mathbf{y}
- Physical Law: $d\mathbf{y}/dt = \mathbf{h}(\mathbf{y}, t)$
- Initial Condition: $\mathbf{y}(t_0) = \mathbf{y}_0$



Atmospheric Models

- \mathbf{X} is the prediction of \mathbf{Y}
- $d\mathbf{X}/dt = \mathbf{f}(\mathbf{X}, t) + q(t)\mathbf{X}$
- Initial Condition: $\mathbf{X}(t_0) = \mathbf{X}_0$
- Stochastic Forcing:
 - $\langle q(t) \rangle = 0$
 - $\langle q(t)q(t') \rangle = q^2\delta(t-t')$



Model Error



- $Z = X - Y$

- Initial: $Z_0 = X_0 - Y_0$

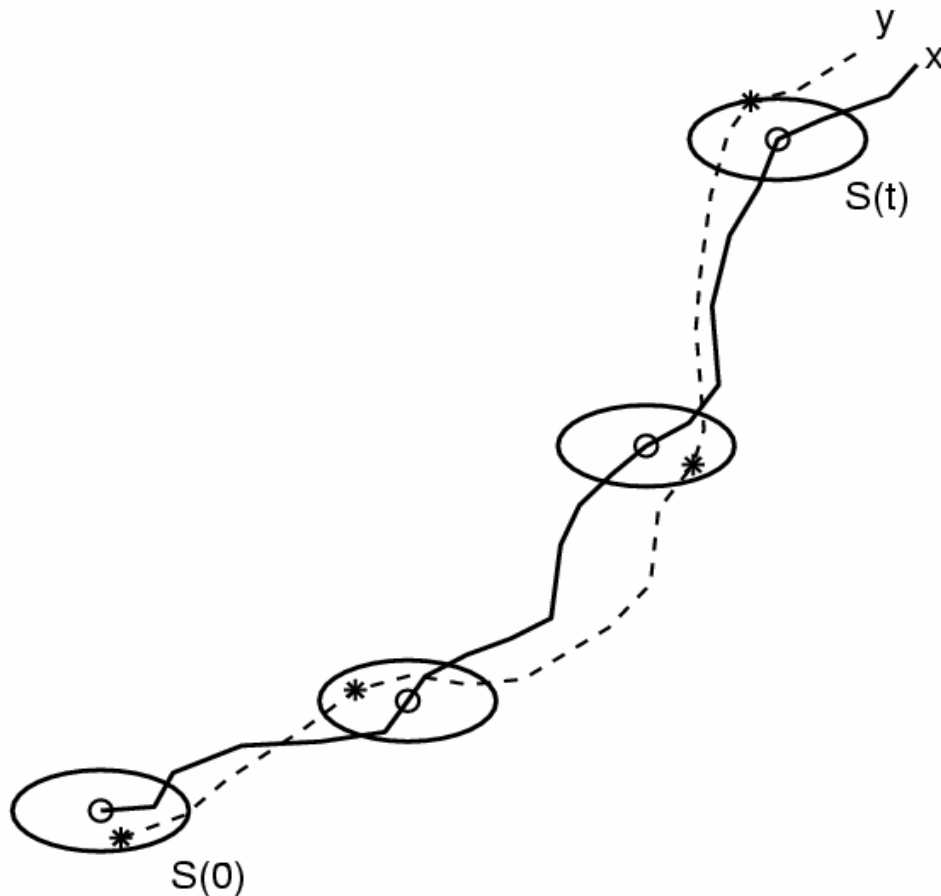


One Overlooked Parameter

- Tolerance Level ε
- Maximum accepted error



Valid Predict Period (VPP)



- VPP is defined as the time period when the prediction error first exceeds a pre-determined criterion (i.e., the tolerance level ε).



First-Passage Time





Conditional Probability Density Function

- Initial Error: \mathbf{z}_0
- $(t - t_0)$ ↗ Random Variable
- Conditional PDF of $(t - t_0)$ with given \mathbf{z}_0 ↗
 - $P[(t - t_0) | \mathbf{z}_0]$



Two Approaches to Obtain PDF of VPP

- Analytical (Backward Fokker-Planck Equation)
- Practical



Backward Fokker-Planck Equation

$$\frac{\partial P}{\partial t} - [\mathbf{f}(\mathbf{z}_0, t)] \frac{\partial P}{\partial \mathbf{z}_0} - \frac{1}{2} q^2 \mathbf{z}_0^2 \frac{\partial^2 P}{\partial \mathbf{z}_0 \partial \mathbf{z}_0} = 0$$



Moments

$$\tau_1(\mathbf{z}_0) = \int_{t_0}^{\infty} P(t_0, \mathbf{z}_0, t - t_0)(t - t_0) dt$$

$$\tau_2(\mathbf{z}_0) = \int_{t_0}^{\infty} P(t_0, \mathbf{z}_0, t - t_0)(t - t_0)^2 dt$$



Example: Maximum Growing Manifold of Lorenz System (Nicolis, 1992)



$$\frac{d\xi}{dt} = (\sigma - g\xi^2) + v(t)\xi, \quad 0 \leq \xi < \infty$$

$$\langle v(t) \rangle = 0, \quad \langle v(t)v(t') \rangle = q^2 \delta(t - t').$$

$$\sigma = 0.64, \quad g = 0.3, \quad q^2 = 0.2.$$



Mean and Variance of VPP

$$(\sigma\xi_0 - g\xi_0^2) \frac{d\tau_1}{d\xi_0} + \frac{q^2\xi_0^2}{2} \frac{d^2\tau_1}{d\xi_0^2} = -1$$

$$(\sigma\xi_0 - g\xi_0^2) \frac{d\tau_2}{d\xi_0} + \frac{q^2\xi_0^2}{2} \frac{d^2\tau_2}{d\xi_0^2} = -2\tau_1$$

$$\tau_1 = 0, \quad \tau_2 = 0 \quad \text{for } \xi_0 = \varepsilon.$$

$$\frac{d\tau_1}{d\xi_0} = 0, \quad \frac{d\tau_2}{d\xi_0} = 0 \quad \text{for } \xi_0 = \xi_{\text{noise}}.$$



Analytical Solutions

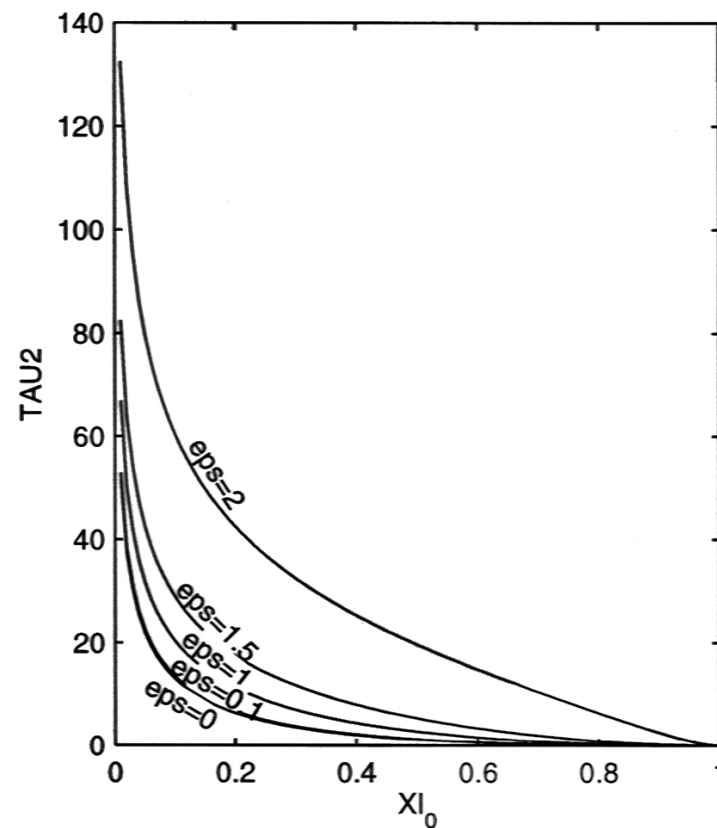
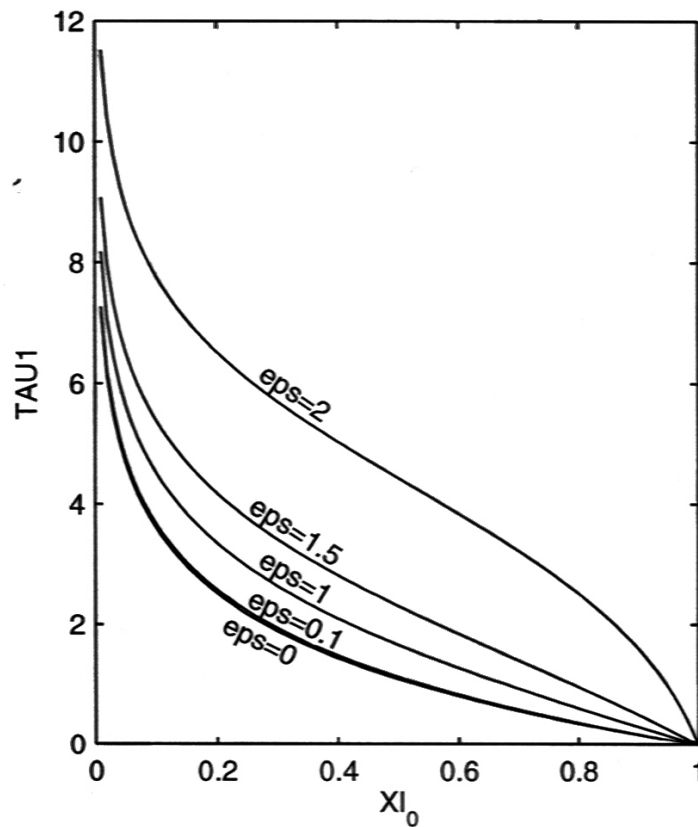
$$\tau_1(\bar{\xi}_0, \bar{\xi}_{noise}, \varepsilon) = \frac{2}{q^2} \int_{\bar{\xi}_0}^1 y^{-\frac{2\sigma}{q^2}} \exp\left(\frac{2\varepsilon g}{q^2} y\right) \left[\int_{\bar{\xi}_{noise}}^y x^{\frac{2\sigma}{q^2}-2} \exp\left(-\frac{2\varepsilon g}{q^2} x\right) dx \right] dy$$

$$\tau_2(\bar{\xi}_0, \bar{\xi}_{noise}, \varepsilon) = \frac{4}{q^2} \int_{\bar{\xi}_0}^1 y^{-\frac{2\sigma}{q^2}} \exp\left(\frac{2\varepsilon g}{q^2} y\right) \left[\int_{\bar{\xi}_{noise}}^y \tau_1(x) x^{\frac{2\sigma}{q^2}-2} \exp\left(-\frac{2\varepsilon g}{q^2} x\right) dx \right] dy$$

$$\bar{\xi}_0 = \xi_0 / \varepsilon,$$

$$\bar{\xi}_{noise} = \xi_{noise} / \varepsilon$$

Dependence of tau1 & tau2 on Initial Condition Error (ξ_0/ε)





Practical Approach

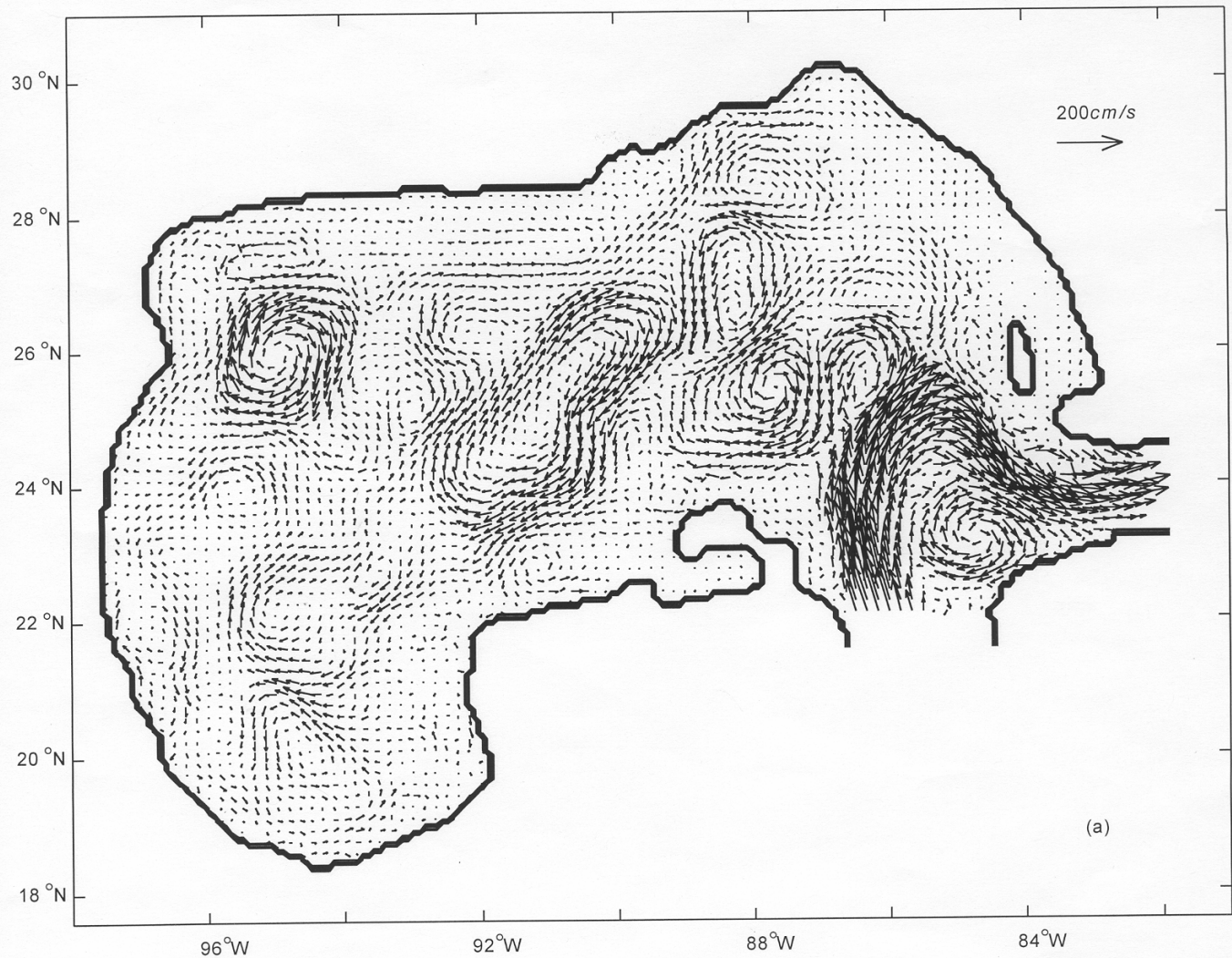
Gulf of Mexico Ocean Prediction
System



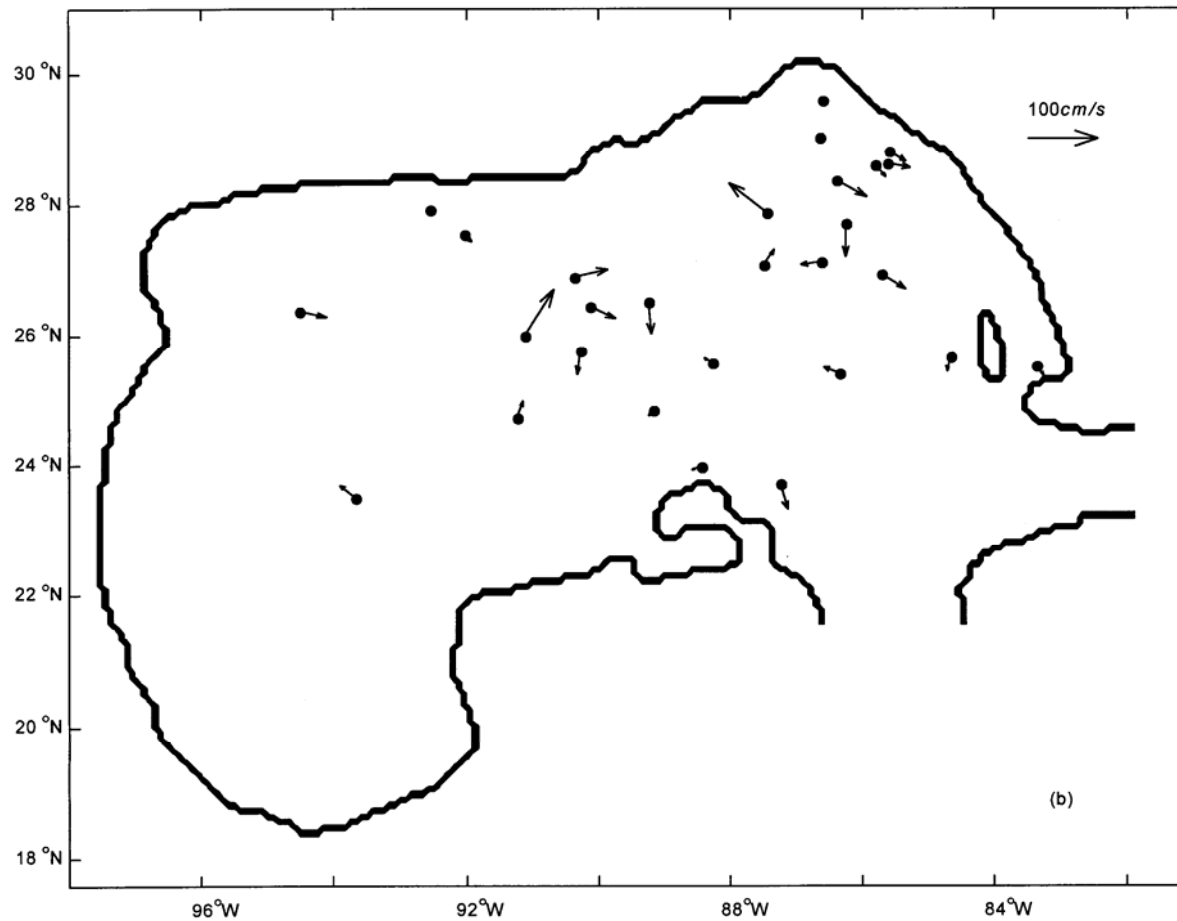
Gulf of Mexico Forecast System

- University of Colorado Version of POM
- $1/12^\circ$ Resolution
- Real-Time SSH Data (TOPEX, ESA ERS-1/2) Assimilated
- Real Time SST Data (MCSST, NOAA AVHRR) Assimilated
- Six Months Four-Times Daily Data From July 9, 1998 for Verification

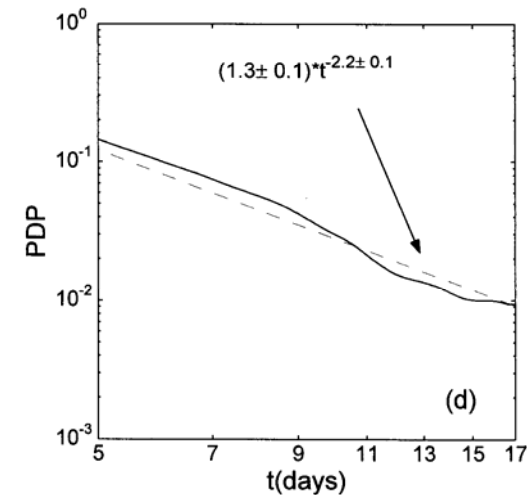
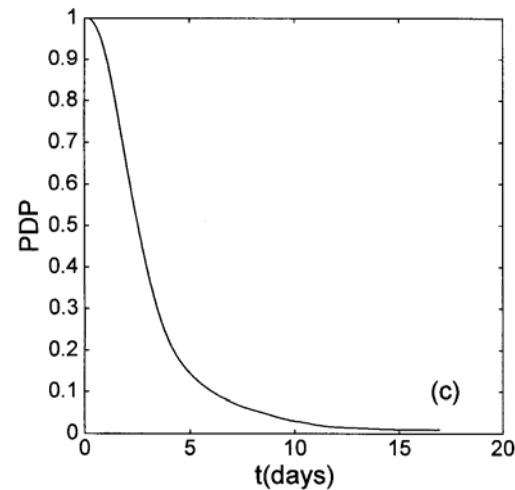
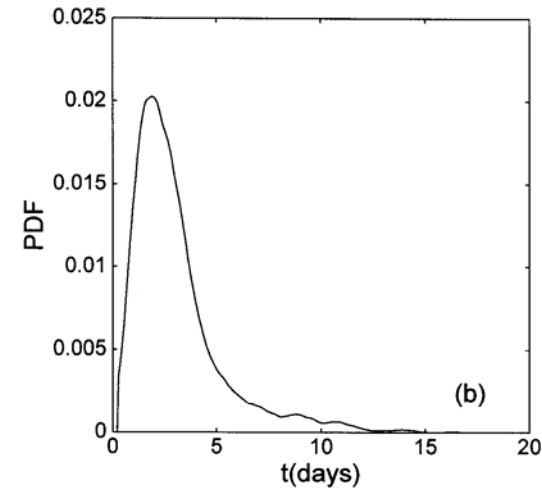
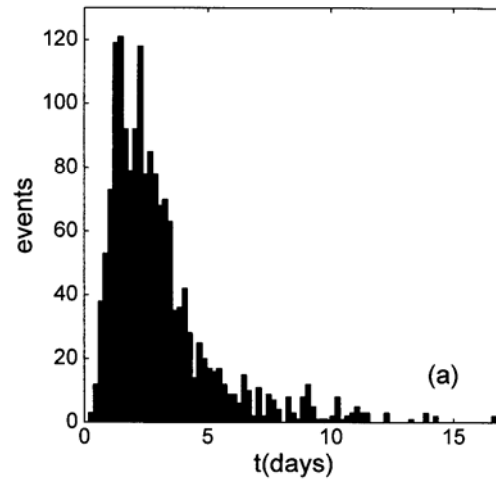
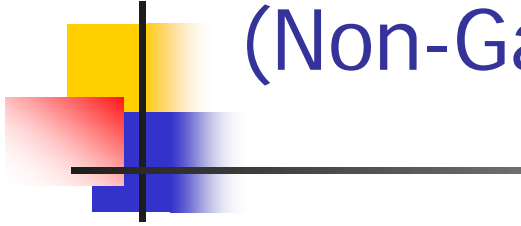
Model Generated Velocity Vectors at 50 m on 00:00 July 9, 1998



(Observational) Drifter Data at 50 m on 00:00 July 9, 1998



Statistical Characteristics of VPP for zero initial error and 22 km tolerance level (Non-Gaussian)





Conclusions

- (1) VPP (i.e., FPT) is an effective prediction skill measure (scalar).
- (2) Theoretical framework for FPT (such as Backward Fokker-Planck equation) can be directly used for model predictability study.