

The Sensitivity of Adjoint Sensitivity

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The Problem

Adjoint sensitivity is useful for a variety of applications, but exhibits significant variability with respect to various parameters such as:

- Model resolution
- Basic-state trajectory
- Model physics

The goal of this study is to understand the “sensitivity of sensitivity” toward the best use of sensitivity in data assimilation, observation targeting, and dynamical interpretation applications.

Sensitivity Variability Background

Adjoint sensitivity formulation for a forecast response function J:

$$\frac{\partial J}{\partial \mathbf{Y}_0} = \left(\frac{\partial \mathbf{f}(\mathbf{Y})}{\partial \mathbf{Y}_{0,t}} \right)^T * \frac{\partial J}{\partial \mathbf{Y}_t} \quad (1)$$

Adjoint Sensitivity variables Tangent-linear propagator Differentiated J w.r.t. forecast variables

- $d\mathbf{Y}/dt = \mathbf{f}(\mathbf{Y})$ is the nonlinear forward forecast model
- \mathbf{Y} is a vector of model state variables

The above formula represents the adjoint sensitivity field about a single, previously-run forecast begun from initial condition \mathbf{Y}_0 .

The adjoint sensitivity about a 2nd forecast trajectory ($\partial J / \partial \mathbf{Y}_0'$) begun from initial condition $\mathbf{Y}_0 + \Delta \mathbf{Y}_0$ can be approximated by:

$$\frac{\partial J}{\partial \mathbf{Y}_0'} = \left(\frac{\partial \mathbf{f}(\mathbf{Y}_0)}{\partial \mathbf{Y}_{0,t}} + \frac{\partial^2 \mathbf{f}(\mathbf{Y}_0)}{\partial \mathbf{Y}_{0,t}^2} * \Delta \mathbf{Y}_0 \right)^T * \frac{\partial J}{\partial \mathbf{Y}_t} \quad (2)$$

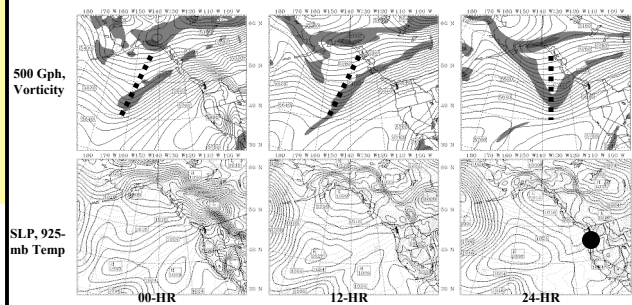
Subtracting (2) from (1) yields the difference in adjoint sensitivity calculated about 2 different forecast trajectories:

$$\Delta \frac{\partial J}{\partial \mathbf{Y}_0} = \left(\frac{\partial^2 \mathbf{f}(\mathbf{Y}_0)}{\partial \mathbf{Y}_{0,t}^2} * \Delta \mathbf{Y}_0 \right)^T * \frac{\partial J}{\partial \mathbf{Y}_t} \quad (3)$$

- Linear $\mathbf{f}(\mathbf{Y})$ gives same sensitivity for all forecasts ($\Delta \frac{\partial J}{\partial \mathbf{Y}_0} = 0$)
- $\Delta \frac{\partial J}{\partial \mathbf{Y}_0}$ for Nonlinear $\mathbf{f}(\mathbf{Y})$ depends on initial condition differences ($\Delta \mathbf{Y}_0$) and the forecasts themselves (for nonlinearity greater than quadratic)

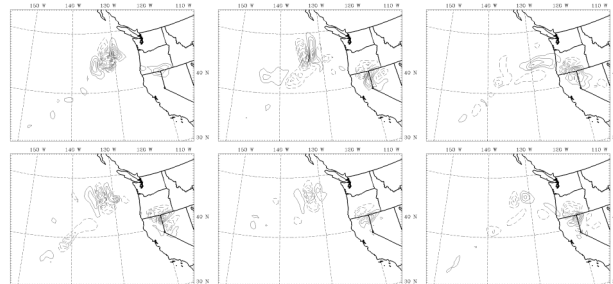
So how do sensitivities vary in practical forecasting applications?

The following plots illustrate a typical wintertime synoptic pattern impacting the Pacific Northwest U.S.:



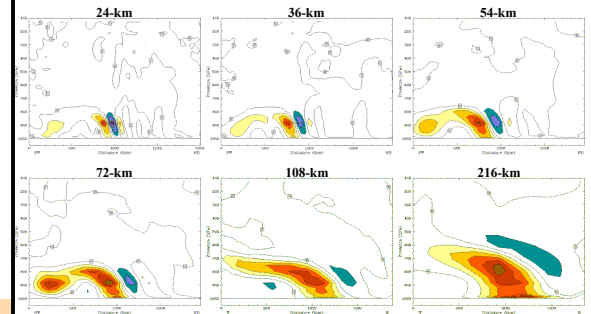
In this scenario, a trough aloft moves toward the North American coastline, forcing a weak cyclone into the PNW at 24-hr forecast time.

Using a **single-point SLP response function at 24-hr** at the black dot above, the adjoint sensitivity was calculated about all 90 equally-likely forecasts of the University of Washington Ensemble Kalman Filter (EnKF) for this case. The following plots represent the **sensitivity to 00-hr 850-mb temperature** calculated about 6 of these forecasts:



- Significant differences exist in structure, location, and magnitude among the 6 equally-likely initial-condition sensitivity fields
- These differences result in a large range of predicted response function perturbations for identical initial-condition perturbations
- Nonlinear perturbation evolution about the EnKF mean in this case is to blame for these discrepancies, and reveals the difficulty in diagnosing mesoscale regions where errors can grow rapidly

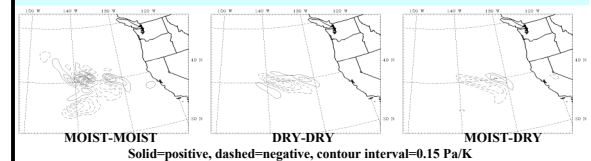
Significant differences in sensitivity also arise when calculating sensitivity at different horizontal resolutions. For a similar synoptic situation and 24-hr SLP response function, the following plots depict vertical cross-sections of sensitivity to 00-hr temperature:



Warm colors=negative, cool colors=positive, contour interval varies

- Differences are apparent in the horizontal scale, degree of vertical tilt, level of maximum sensitivity, and vertical extent of the adjoint sensitivity fields
- The sensitivity field magnitudes increase at higher resolution and are about 50 times larger at 24-km resolution than at 216-km resolution

Differences also exist using simplified adjoint physics:



- Simplified adjoint physics (dry in this case) appear to be the limiting factor, causing sensitivity to resemble that of the totally dry case, even with a moist forward trajectory

Conclusions

- Much like an ensemble of forecasts, adjoint sensitivity should be thought of probabilistically
- Significant differences in the sensitivity field can occur using different model parameters like resolution or physics, and these issues should be considered as possible sources of error in various adjoint applications