UVIG Tutorial on Integration of Uncertainty Forecasts into the Power System Operations

Part 1: Background, Methods, and Meaning of Uncertainty Forecasts

Overview of Techniques and Issues Associated with Probabilitstic Forecasting

Sue Ellen Haupt

National Center for Atmospheric Research Research Applications Laboratory

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The Plan

- Why do we need probabilistic forecasts?
- How do we construct ensembles?
- How can we tell if the ensemble is "good"?
- How can we make an imperfect ensemble better?
- How can we extract information from uncertainty information from forecasts?
- What else do we need to know?

We wish to predict specific events

8/03/09 771mw up-ramp from 20:10 - 22:10 followed by a 738mw down-ramp from 22:40 - 00:50



Time (LST)

Why is Atmospheric Flow Subject to Uncertainy?



Ed Lorenz



- Nonlinearity
- Sensitivity to initial conditions
- Chaos → There are limits to predictability
- Think in terms of attractors & manifolds
- Requires probabilistic forecasts

Ensembles & Uncertainty Quantification

- Account for uncertainties due to imperfect initial conditions and model formulation
- Produce more accurate predictions than any single model realization
- Provide flowdependent uncertainty estimates



Short Range Ensemble Forecast System

30 hr forecast



10 m Wind 500 mb Height

How are Ensembles Generated?

- Perturb initial conditions
- Different boundary conditions
- Include different physics
- Different models

How Many Members are Needed?

Good spread-error correlation only with very large ensembles (order hundreds). From Kolczynski et al. (2011, MWR)).

Can we reduce that number?

- 10 member ensemble with nearly same CRPS value as 42 member ensemble
- Lower CRPS with calibration (Bayesian Model Averaging)

What if we had only one member? Analog Prediction

Spread-skill relathionship

Luca Delle Monache

How do we determine a Good Match?

• Since probabilistic, need to evaluate based on large number of forecasts Spread-skill

The Brier Score

• Mean square error of a probability forecast

$$BS = \frac{1}{n} \sum_{i=1}^{n} (f_i - x_i)^{n}$$

n is the number of forecasts where f_i is the forecast prob on occasion i x_i is the observation (0 or 1) on occasion *i*

Resolution

Distinguish

different events

• Weights larger errors more than smaller ones

0.3

Reliability

Frequency

matches actual

Jared Lee

Variability of

Observations

Continuous Ranked Probability Score

How do we improve the Match?

- Centered (the right answer)
- Sharp (narrow range)
- Reliable (quantile predicted matches quantile observed averaged over time)

Example Calibration

Example Calibration Techniques

• Linear Variance Calibration

- Ensemble Kalman Filter
- Quantile Regression
- Bayesian Model Averaging
- Kernal Density Methods
- Analogue Method
- Many others, including logistic regression, nonhomogeneous Gaussian regression, EMOS,

Aviation decision making—constraint

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Matt Wandishin

Comparing the Two Strategies Over Time

3TIER P70 and 50% scaled forecast similar risk exposure (27% vs. 28%) 3TIER P70 scheduled 20.5 GWh more energy than scaled forecast! *Reliable risk and more energy scheduled, day-ahead

Eric Grimit

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Making Actionable Decisions From Probabilistic Forecasts

[–] Erik Ela

Summary

The atmosphere is inherently Chaotic

- Ensemble prediction embraces and quantifies the uncertainty, producing
 - > Better mean forecasts
 - Estimates of uncertainty
- >The ensemble should be calibrated
- Research is showing
 - Better ways of creating ensembles
 - Better ways of blending ensemble information via postprocessing

Such probabilistic forecasts can enhance decision-making

