

Model errors for MetUM winds – can we estimate these well with dense observations?

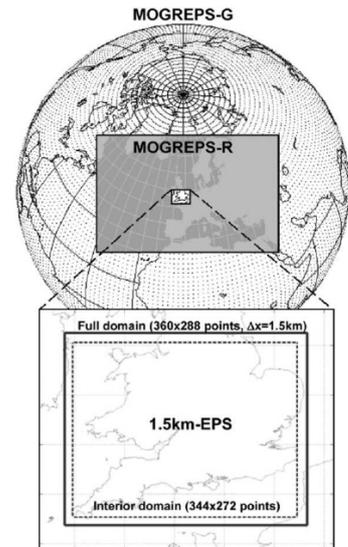
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- The Met Office has a high resolution MetUM over the UK.
- Little is known about the statistics of its errors.
- Diagnosing model error statistics is difficult!

There are different aspects to errors in general:

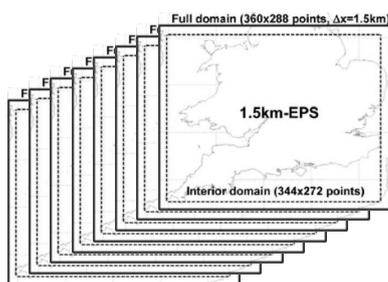
- They can have **random** and **systematic** components.
- They may be **additive** or **multiplicative**
- They may be **Gaussian** or **non-Gaussian**.
- They may be **homogeneous** or **inhomogeneous**.
- They may be **isotropic** or **anisotropic**.
- They may be **correlated** or **uncorrelated**.



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We have for one day (20/09/11):

- 1.8 million Doppler radar observations (Chilbolton Observatory).
- High-resolution forecasts (1.5km grid-length) over the Southern UK.
- A 93-member ensemble of forecasts.



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Assumptions:

	Random only?	Additive?	Gaussian?	Homogeneous?	Isotropic?	Mutually uncorrelated?	Cross correlated?
Model errors	✓	✓	✓	✓	✓	✗	✗
Predictability errors	✓	✓	✓	✓	✓	✗	✗
Observation errors	✓	✓	✓	✓	✓	✓	✗

... plus that the ensemble is correctly spread, and that observations can be treated as 'point-like'. These assumptions are not always justified.

General definitions

Forecast,
analysis, or
observation = 'Truth' + Error realisation

Innovation = Observation - Forecast 'observation'

Covariance = \langle (Quantity 1 - mean of $\mathcal{Q}1$) \times (Quantity 2 - mean of $\mathcal{Q}2$) \rangle

Specific definitions

Forecast (at time t)

$$\mathbf{x}_t^f = \mathbf{x}_t^t + \boldsymbol{\epsilon}_t^f$$

Analysis (at $t = 0$)

$$\mathbf{x}_0^a = \mathbf{x}_0^t + \boldsymbol{\epsilon}_0^a$$

Observations

$$\mathbf{y}_t = \mathcal{H}_t(\mathbf{x}_t^t) + \boldsymbol{\epsilon}_t^{\text{ob}}$$

Observation operator applied to forecast

$$\begin{aligned} \mathcal{H}_t(\mathbf{x}_t^f) &= \mathcal{H}_t(\mathbf{x}_t^t + \boldsymbol{\epsilon}_t^f) \\ &\approx \mathcal{H}_t(\mathbf{x}_t^t) + \mathbf{H}_t \boldsymbol{\epsilon}_t^f \end{aligned}$$

Forecast model (no stochastic scheme)

$$\mathcal{M}(\mathbf{x}_{t-1}^t) = \mathbf{x}_t^t + \boldsymbol{\eta}_t$$

Using the definitions

Applying the forecast model

$$\begin{aligned}
 \mathbf{x}_t^f &= \mathcal{M}_t(\mathbf{x}_{t-1}^a) \\
 &= \mathcal{M}_t(\mathbf{x}_{t-1}^t + \boldsymbol{\epsilon}_{t-1}^a) \\
 &\approx \mathcal{M}_t(\mathbf{x}_{t-1}^t) + \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a \\
 \mathbf{x}_t^t + \boldsymbol{\epsilon}_t^f &= \mathbf{x}_t^t + \underbrace{\boldsymbol{\eta}_t}_{\text{model error}} + \underbrace{\mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a}_{\text{predictability error}}
 \end{aligned}$$

Innovations

$$\begin{aligned}
 \mathbf{d}_t^{o,b} &= \mathbf{y}_t - \mathcal{H}_t(\mathbf{x}_t^f) \\
 &\approx \mathcal{H}_t(\mathbf{x}_t^t) + \boldsymbol{\epsilon}_t^{ob} - \mathcal{H}_t(\mathbf{x}_t^t) - \mathbf{H}_t \boldsymbol{\epsilon}_t^f \\
 &= \boldsymbol{\epsilon}_t^{ob} - \mathbf{H}_t \boldsymbol{\epsilon}_t^f
 \end{aligned}$$

Ensemble perturbations

$$\begin{aligned}
 \boldsymbol{\delta}_t^f &\approx \mathbf{x}_t^f - \langle \mathbf{x}_t^f \rangle \\
 &= \mathcal{M}_t(\mathbf{x}_{t-1}^a) - \langle \mathcal{M}_t(\mathbf{x}_{t-1}^a) \rangle \\
 &= \mathcal{M}_t(\mathbf{x}_{t-1}^t + \boldsymbol{\epsilon}_{t-1}^a) - \langle \mathcal{M}_t(\mathbf{x}_{t-1}^t + \boldsymbol{\epsilon}_{t-1}^a) \rangle \\
 &\approx \mathcal{M}_t(\mathbf{x}_{t-1}^t) + \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a - \langle \mathcal{M}_{t-1 \rightarrow t}(\mathbf{x}_{t-1}^t) \rangle \\
 &= \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a.
 \end{aligned}$$

Procedure – interpretation of Daley (1992)

Daley, Roger (1992) The effect of serially correlated observation and model error on atmospheric data assimilation, Monthly weather review 120 (1), 164 – 177.

1. Analysis of innovations

$$\begin{aligned}
 \langle \mathbf{d}_t^{o,b} \mathbf{d}_t^{o,bT} \rangle &= \langle (\mathbf{y}_t - \mathcal{H}_t(\mathbf{x}_t^f)) (\mathbf{y}_t - \mathcal{H}_t(\mathbf{x}_t^f))^T \rangle \\
 &= \langle (\boldsymbol{\epsilon}_t^{ob} - \mathbf{H}_t \boldsymbol{\epsilon}_t^f) (\boldsymbol{\epsilon}_t^{ob} - \mathbf{H}_t \boldsymbol{\epsilon}_t^f)^T \rangle \\
 &= \langle (\boldsymbol{\epsilon}_t^{ob} - \mathbf{H}_t \boldsymbol{\eta}_t - \mathbf{H}_t \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a) (\boldsymbol{\epsilon}_t^{ob} - \mathbf{H}_t \boldsymbol{\eta}_t - \mathbf{H}_t \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a)^T \rangle \\
 &\approx \underbrace{\langle \boldsymbol{\epsilon}_t^{ob} \boldsymbol{\epsilon}_t^{obT} \rangle}_{\mathbf{R}_t} + \underbrace{\mathbf{H}_t \langle \boldsymbol{\eta}_t \boldsymbol{\eta}_t^T \rangle \mathbf{H}_t^T}_{\mathbf{Q}_t} + \underbrace{\mathbf{H}_t \mathbf{M}_t \langle \boldsymbol{\epsilon}_{t-1}^a \boldsymbol{\epsilon}_{t-1}^{aT} \rangle \mathbf{M}_t^T \mathbf{H}_t^T}_{\mathbf{A}_{t-1}}
 \end{aligned}$$

observation error part
model error part
predictability error part

Warning

Analysis of innovations will be susceptible to biases in obs and in model

2. Ensemble analysis (no localisation)

$$\begin{aligned}
 \langle (\mathbf{H}_t [\mathbf{x}_t^f - \langle \mathbf{x}_t^f \rangle]) (\mathbf{H}_t [\mathbf{x}_t^f - \langle \mathbf{x}_t^f \rangle])^T \rangle &\approx \langle (\mathbf{H}_t [\mathbf{x}_t^t + \boldsymbol{\eta}_t + \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a - \langle \mathbf{x}_t^t + \boldsymbol{\eta}_t + \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a \rangle]) \times \\
 &\quad (\mathbf{H}_t [\mathbf{x}_t^t + \boldsymbol{\eta}_t + \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a - \langle \mathbf{x}_t^t + \boldsymbol{\eta}_t + \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a \rangle])^T \rangle \\
 &= \langle (\mathbf{H}_t [\mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a - \langle \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a \rangle]) \times \\
 &\quad (\mathbf{H}_t [\mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a - \langle \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a \rangle])^T \rangle \\
 &\approx \langle (\mathbf{H}_t \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a) (\mathbf{H}_t \mathbf{M}_t \boldsymbol{\epsilon}_{t-1}^a)^T \rangle \\
 &= \underbrace{\mathbf{H}_t \mathbf{M}_t \mathbf{A}_{t-1} \mathbf{M}_t^T \mathbf{H}_t^T}_{\text{predictability error part}}
 \end{aligned}$$

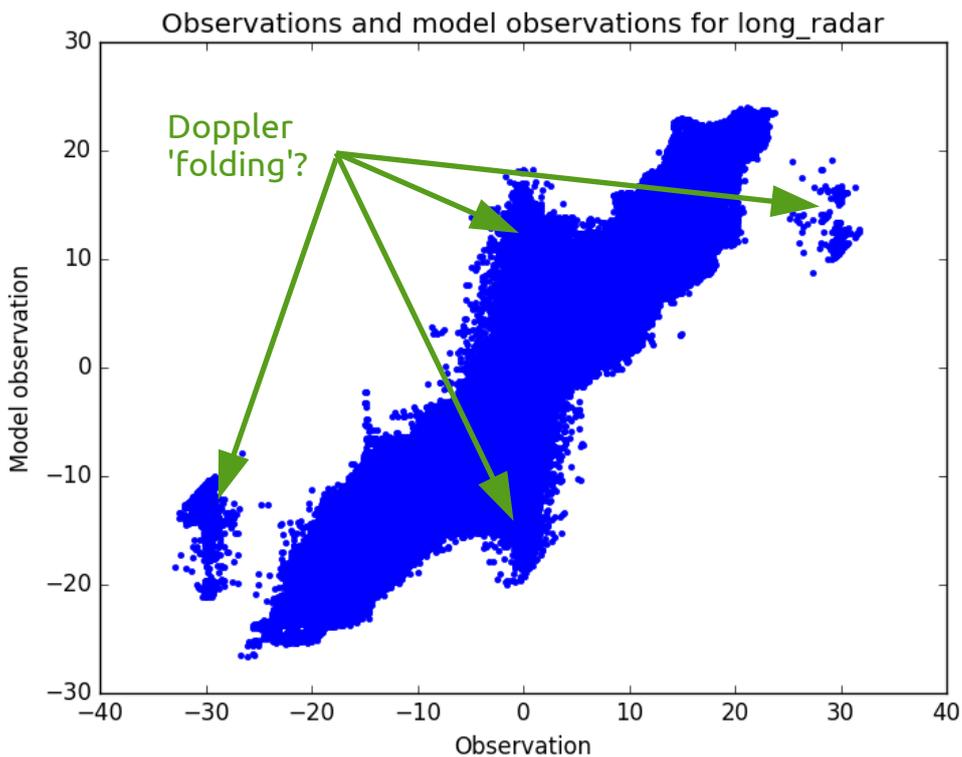
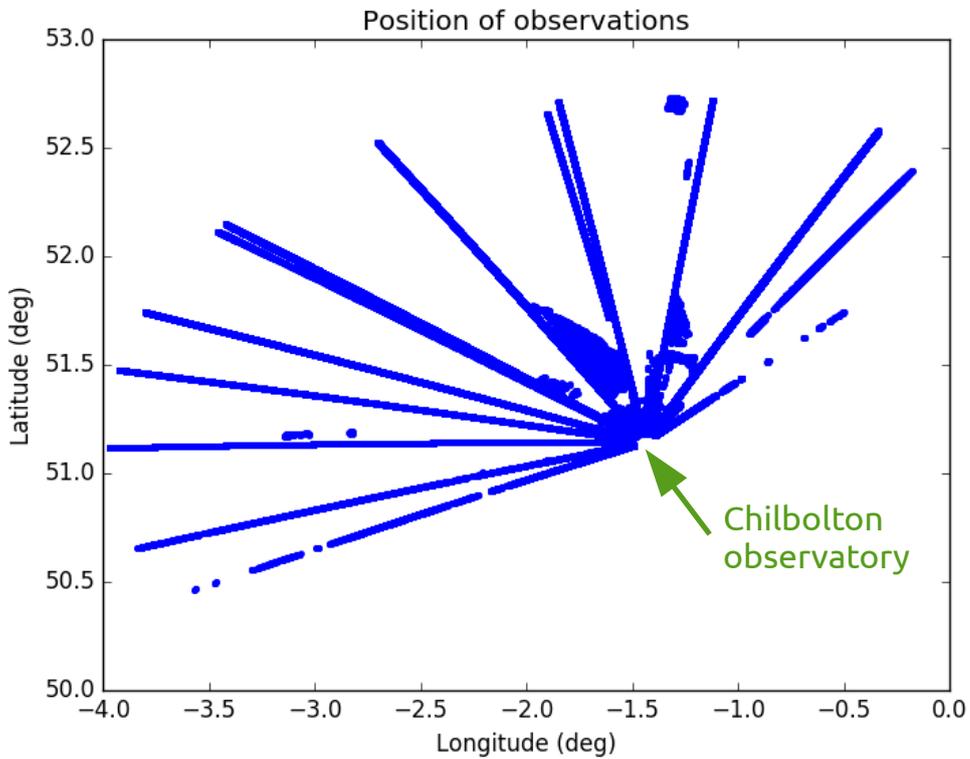
3. Estimate the model error contribution

$$\mathbf{H}_t \langle \boldsymbol{\eta}_t \boldsymbol{\eta}_t^T \rangle \mathbf{H}_t^T \approx \langle \mathbf{d}_t^{o,b} \mathbf{d}_t^{o,bT} \rangle - \langle (\mathbf{H}_t [\mathbf{x}_t^f - \langle \mathbf{x}_t^f \rangle]) (\mathbf{H}_t [\mathbf{x}_t^f - \langle \mathbf{x}_t^f \rangle])^T \rangle - \mathbf{R}_t$$

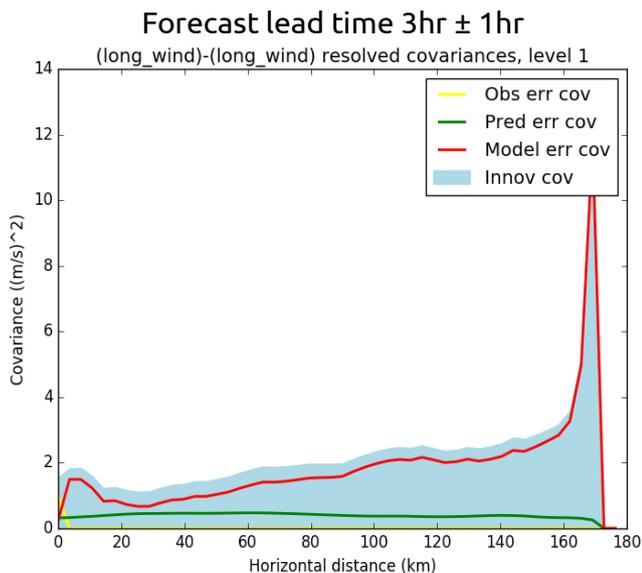
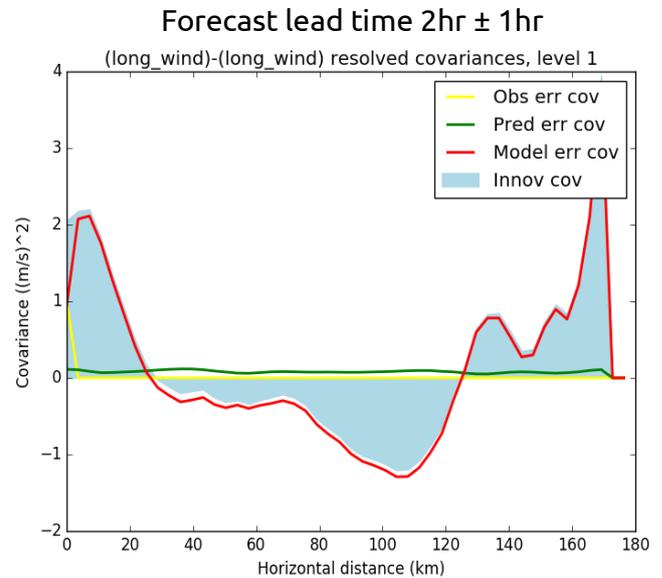
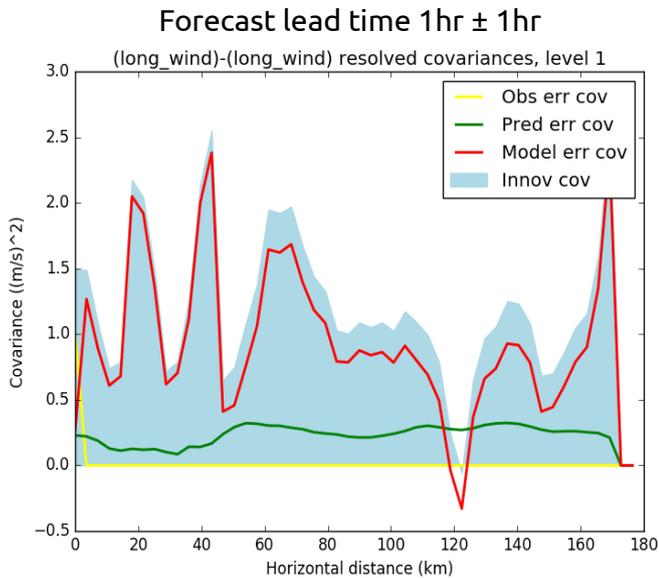
Warning

Model error is estimated as the residual of imperfectly known contributions

The observations and the model's equivalents



Provisional computations



Comments

- No. of samples $\sim O(\text{num obs})^2$.
- Sampling errors are deemed negligible.
- Have assumed contribution from (i) w (vertical wind), and (ii) effect of atmospheric refraction, to 'model obs' are negligible.
- The above results show systematic problems with the procedure on P. 4.
- Some covariances look noisy and increase with distance.
 - Is this an artefact?
 - If not, why does this happen?
- What is the source of the large innovation covariance at ~ 170 km?
- Need to do quality control on observations (e.g. Doppler 'folding')?
- Need to check for systematic errors, and remove where possible.
- How can these results be used?