

**USING SATELLITE-DERIVED ATMOSPHERIC MOTION VECTORS  
(AMV) OBSERVATIONS IN THE DATA ASSIMILATION SYSTEM  
BASED ON LETKF ALGORITHM**

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## TASKS:

- Preparing initial data for SL-AV model;
- Using satellite-derived atmospheric motion vectors (AMV) observations – an important source of wind information in the troposphere with a global coverage.

## DATA AND METHODS USED:

- Data assimilation algorithm – Local Ensemble Transform Kalman Filter (LETKF, Hunt et.al., 2007);
- SL-AV atmospheric model (Tolstykh, 2001) -  $0.9^{\circ} \times 0.72^{\circ}$ , 28 vertical levels;
- Observations used:
  - Ground stations and ships (SYNOP, SHIP: Ps, T2m, RH2m, about 10000 observations)
  - Radiosonde observations (TEMP: T, RH, U, V, about 2000 observations);
  - Aircraft reports (AIREP: T, U, V, about 10000 observations);
  - Atmospheric motion vectors (AMV: U, V, more than 50000 observations)

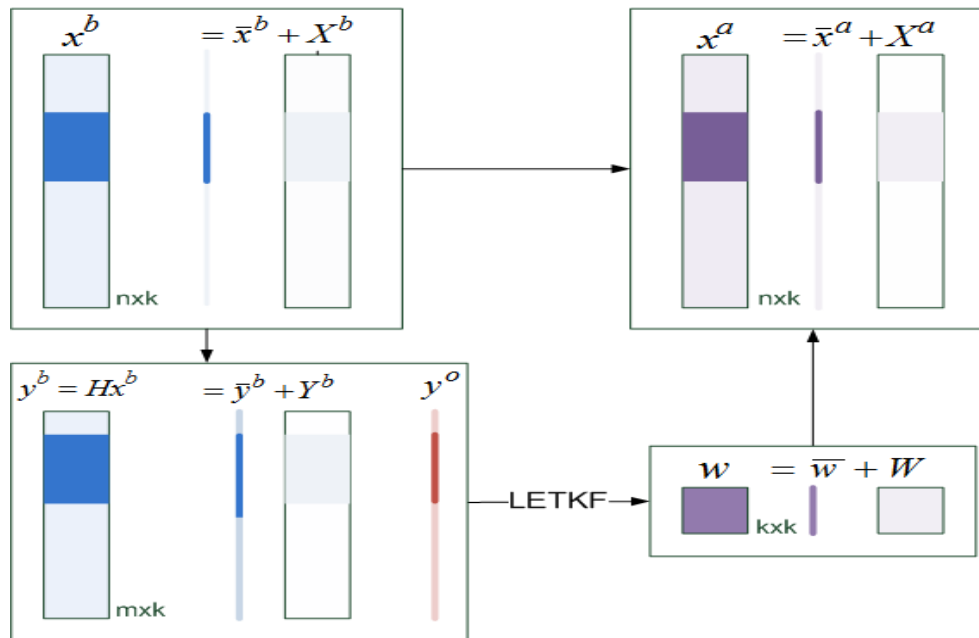
# Local Ensemble Transform Kalman Filter (LETKF)

Local Ensemble Transform Kalman Filter (LETKF, Hunt et.al., 2007):

- square root ensemble filter (no observation perturbations);
- observation localization
- looking for the background ensemble members weights; solving in the ensemble space (usually smaller dimension then the observation, or model space): instead of minimizing  $J(x)$  minimize  $J(w)$

$$J(x) = (x - \bar{x}^b)^T (P^b)^{-1} (x - \bar{x}^b) + (y^o - H(x))^T R^{-1} (y^o - H(x))$$

$$J(w) = (k-1)w^T w + (y^o - \bar{y}^b - Y^b w)^T R^{-1} (y^o - \bar{y}^b - Y^b w)$$



analysis in the model space:

$$P^a = X^b \tilde{P}^a X^{bT}$$

$$\bar{x}^a = \bar{x}^b + X^b \bar{w}^a$$

$$x^{a(i)} = \bar{x}^b + X^b (\bar{w}^a + W^{a(i)})$$

analysis in the ensemble space:

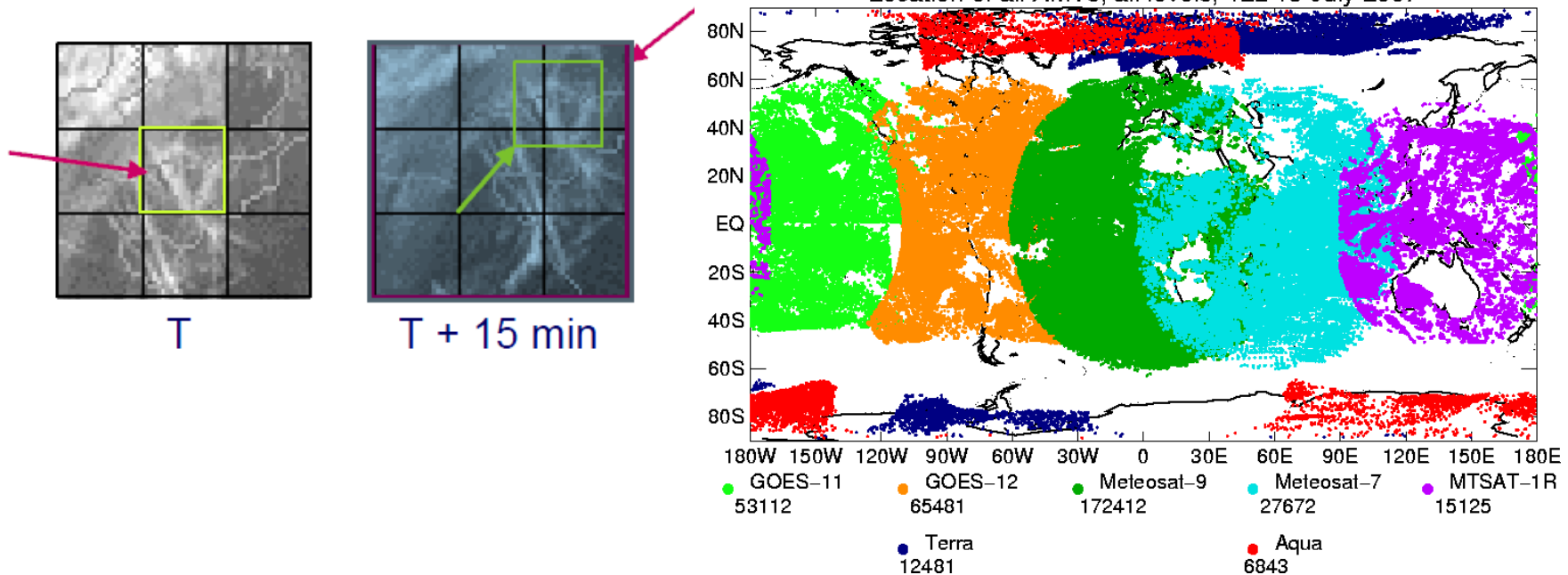
$$\tilde{P}^a = \left( (k-1)I + Y^{bT} R^{-1} Y^b \right)^{-1}$$

$$\bar{w}^a = \tilde{P}^a Y^{bT} R^{-1} (y^o - \bar{y}^b)$$

$$W^a = \left[ (k-1) \tilde{P}^a \right]^{1/2}$$

# Atmospheric Motion Vector (AMV)

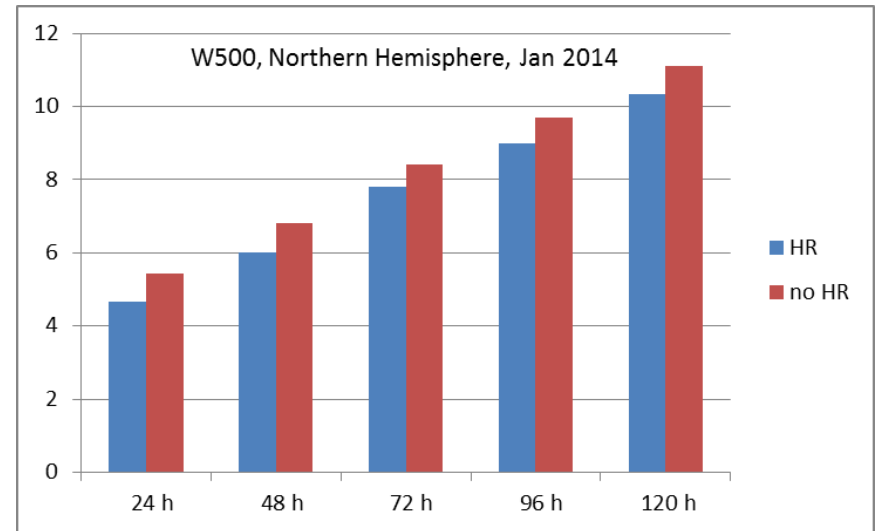
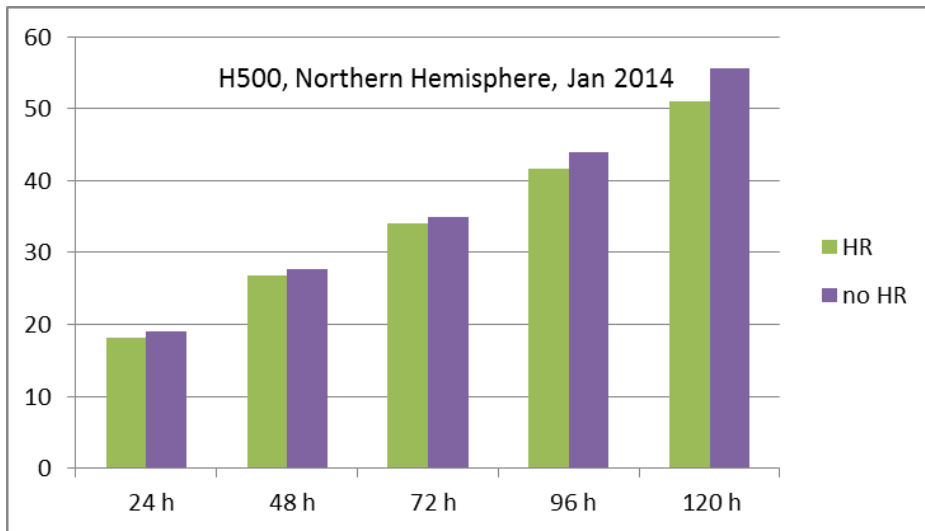
- Produced by tracking clouds or gradients through consecutive satellite images
- Provide the main source of tropospheric wind information over the polar regions.
- Complementary coverage to the geostationary AMV data



- height assignment - largest source of the observation error
- huge number – needs to be thinned (one obs per 200km x 200km x 100hPa box)
- errors are complicated and are spatially and temporally correlated – this necessitates the use of the non-diagonal observation-error covariance matrix  $R$  in the data assimilation scheme

# AMV height reassignment

RMS CBS/WMO Standards, Forecast vs Analysis, Nov 2014, northern, 00 UTC



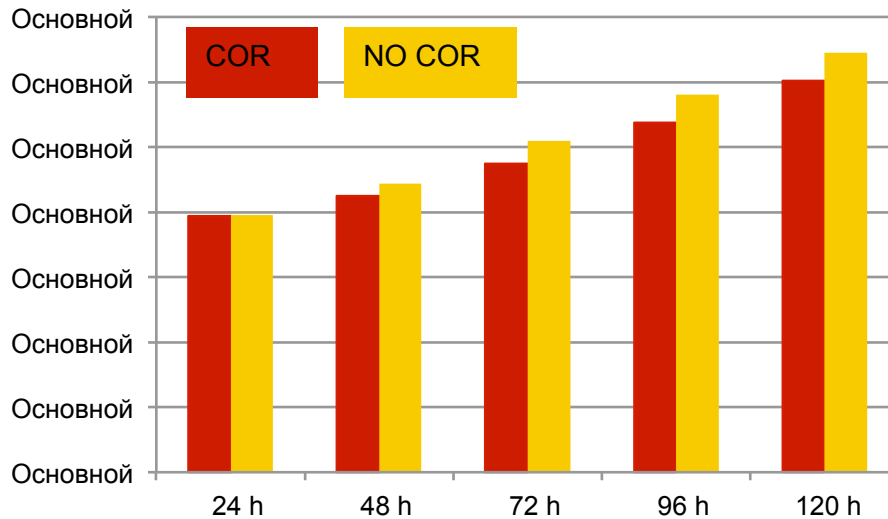
# Accounting of observation errors correlations

- correlation function in the local subset of AMVs is:

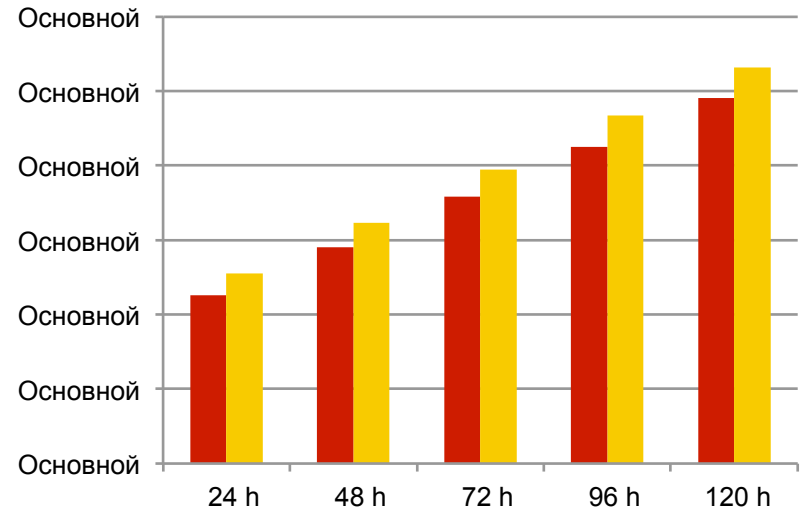
$$R(r) = R_0 \left( 1 + \frac{r}{L} \right)$$

- $R_0$  and  $L$  are different for different satellites, channels and regions
- no cross-variable and cross-obstypes error correlations are assumed

RMS CBS/WMO Standards, Forecast vs Analysis, Nov 2014, northern, 00 UTC, T500



RMS CBS/WMO Standards, Forecast vs Analysis, Nov 2014, northern, 12UTC, W500



# Plans

- Further tuning to reduce errors in CBS/WMO Standards verification
- Use of the parallel I/O to reduce runtime and memory using
- Use of the other observation types
- Regular runs in operational mode
- Use of the filter to generate initial data for medium-range ensemble forecasts

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**THANKS FOR YOUR ATTENTION!**