

# Improvements in Low Level Turbulence (LLT) forecasting for UAVs

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National Center for Atmospheric Research (NCAR)

**3<sup>rd</sup> Turbulence Impact Mitigation Workshop**

**Advances in Detection, Forecasting, and Characterization of Aviation-scale Turbulence**

**5-6 September 2018, MITRE campus, McLean, VA**

**NCAR**



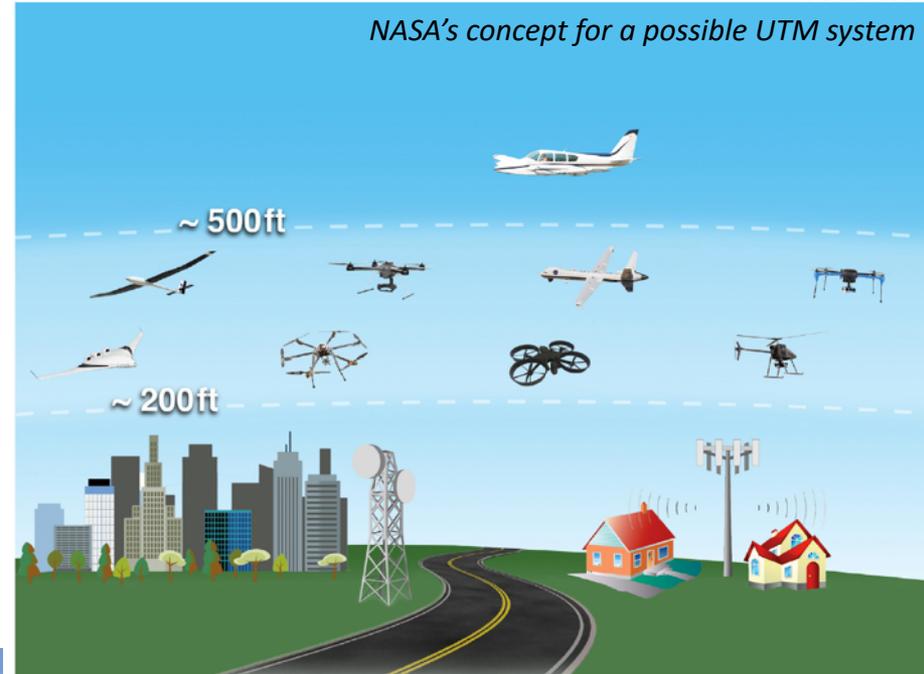
# Unmanned Aerial Systems (UASs)



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Small size/weight of UASs makes them more susceptible to turbulence variability & magnitude [expected low-altitude operations:  $z \approx 50 - 150$  m]

NASA's concept for a possible UTM system



> 10 kg; > 2 m



< 0.1 kg  
< 0.1 m



What is the relevant range of EDR when forecasting turbulence for UAS applications?

# LLT implications for general aviation



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Wake vortices persist longer in weak-turbulence background environments

Require accurate LLT forecasting & have a direct impact in take-off and landing maneuvers



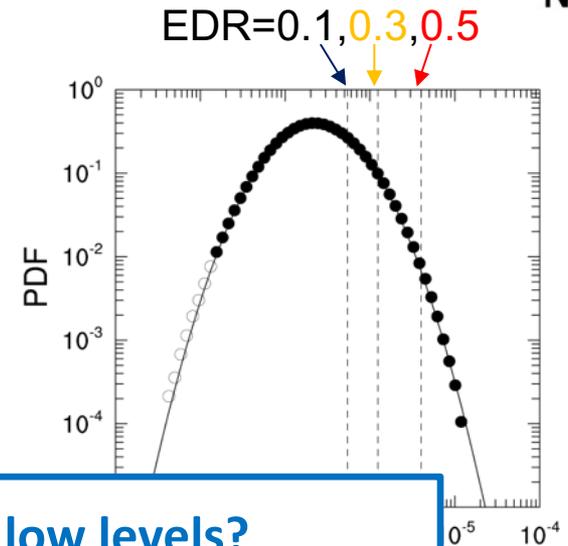
# GTG: “turbulence indices” + “observations”

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Each turbulence index  $D_i$  is rescaled to an EDR assuming a log-normal distribution of EDR

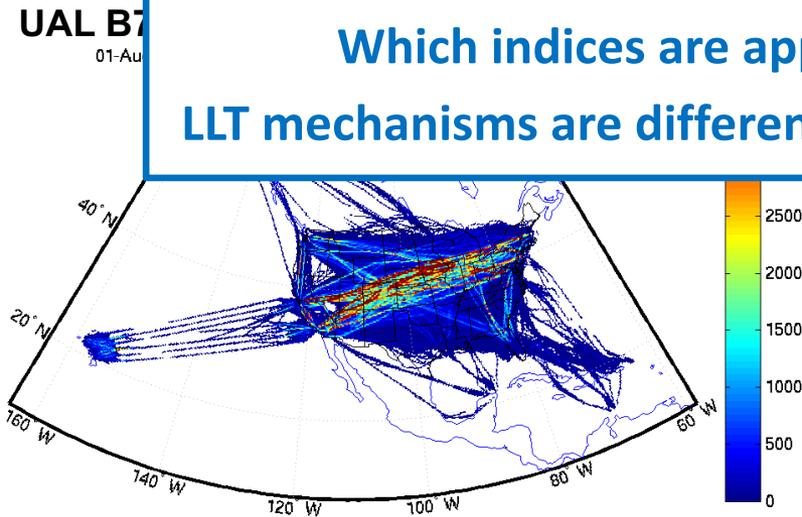
$$\log D_i^* = a + b \log D_i$$

where “a” and “b” are chosen to give best fit to expected log-normal distribution at upper levels and depend on climatology

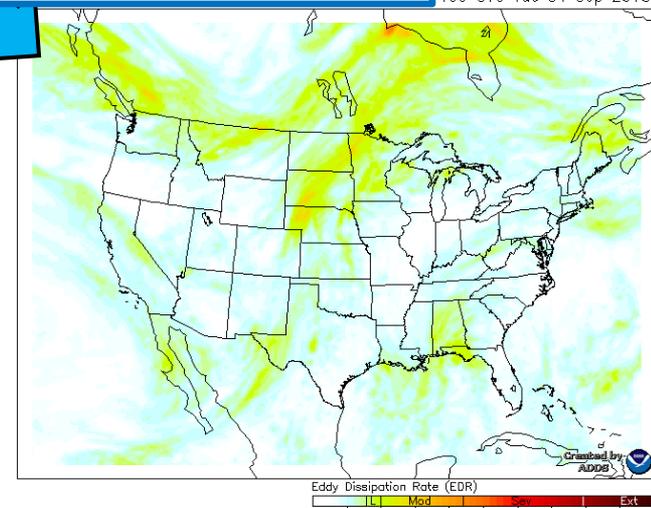


**Which indices are appropriate at low levels?  
LLT mechanisms are different: do we need to modify GTG?**

400 UTC Tue 04 Sep 2018



$n \sum_i w_i D_i^*$



Graphical Turbulence Guidance (GTG)  
Sharman & Pearson (JAMC 2017)  
Pearson & Sharman (JAMC 2017)  
Muñoz-Esparza & Sharman (JAMC 2018)

# EDR observations at low-levels



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The eXperimental Planetary boundary layer Instrumentation Assessment campaign (2015)

(Lundquist et al. BAMS 2017) -> **high-frequency sonic anemometer data**

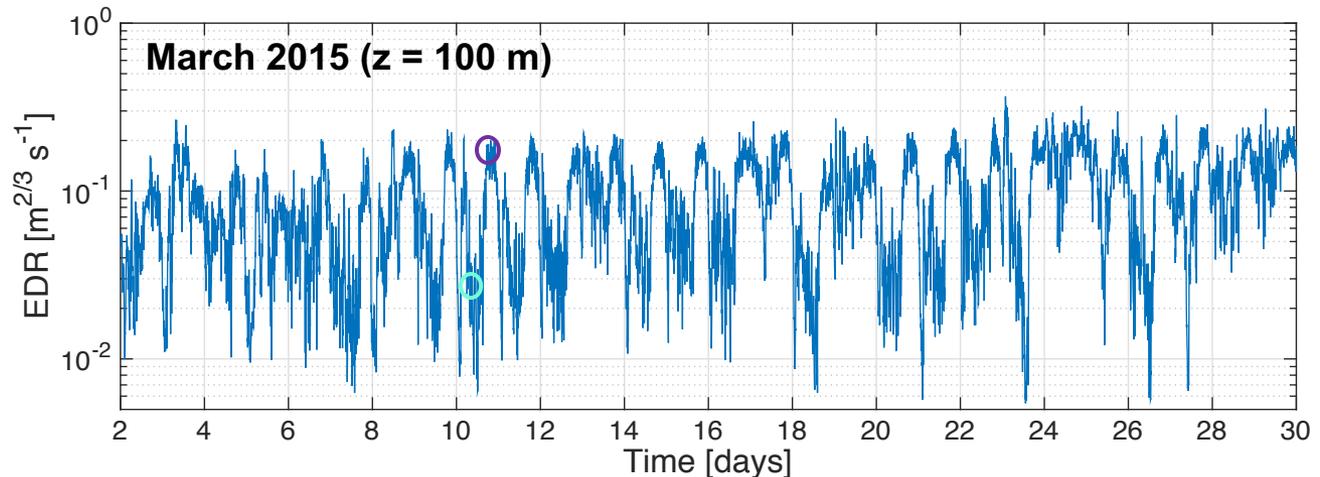
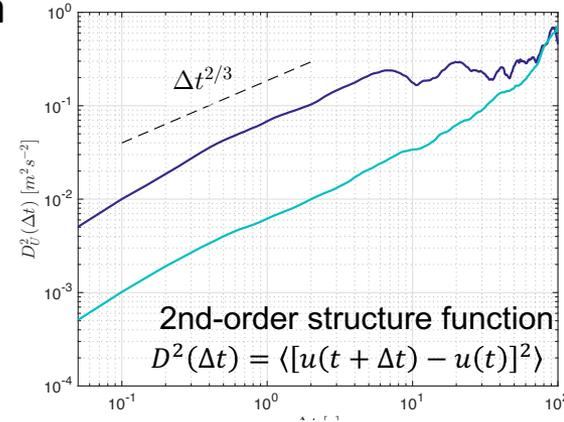


**BAO tower**



Kolmogorov's (1941)  
hypothesis:

$$D^2(\Delta t) = U^{2/3} C_K \varepsilon^{2/3} \Delta t^{2/3}$$



Automated algorithm developed to derive EDRs from sonic anemometers  
**Large dataset generated!!!** (3 months, 7 heights, ~1.5 Million EDRs)

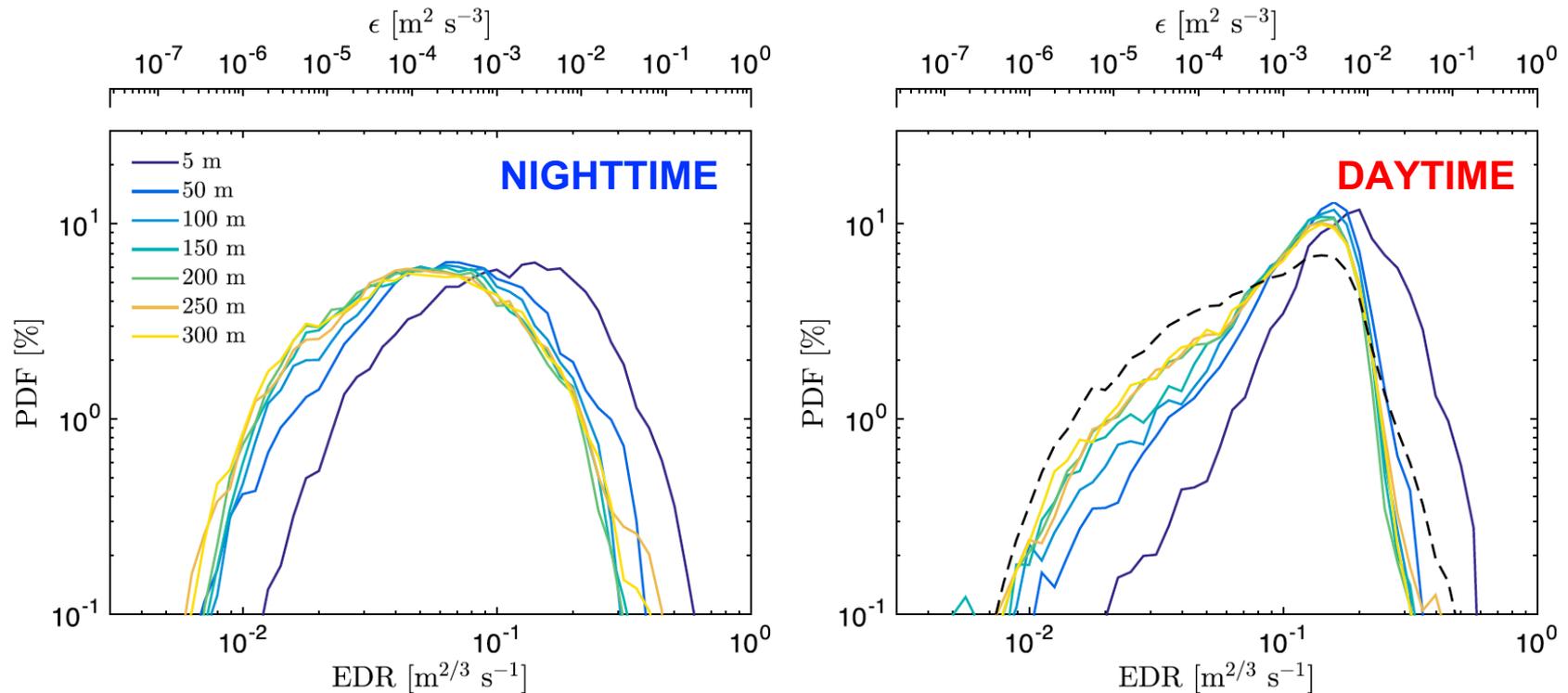
Muñoz-Esparza, Sharman, Lundquist (MWR 2018)

# Insights into low-level turbulence



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## Statistical behavior



Strong diurnal and height dependence

- **Daytime turbulence:** Weibull distribution
- **Nighttime turbulence:** log-normal distribution (same as upper-levels)

# New GTG LLT algorithm & calibration

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Extended statistical re-mapping approach from GTG to ABL stability

(2 distributions: log-normal/Weibull) dependence for low-level EDR forecasting

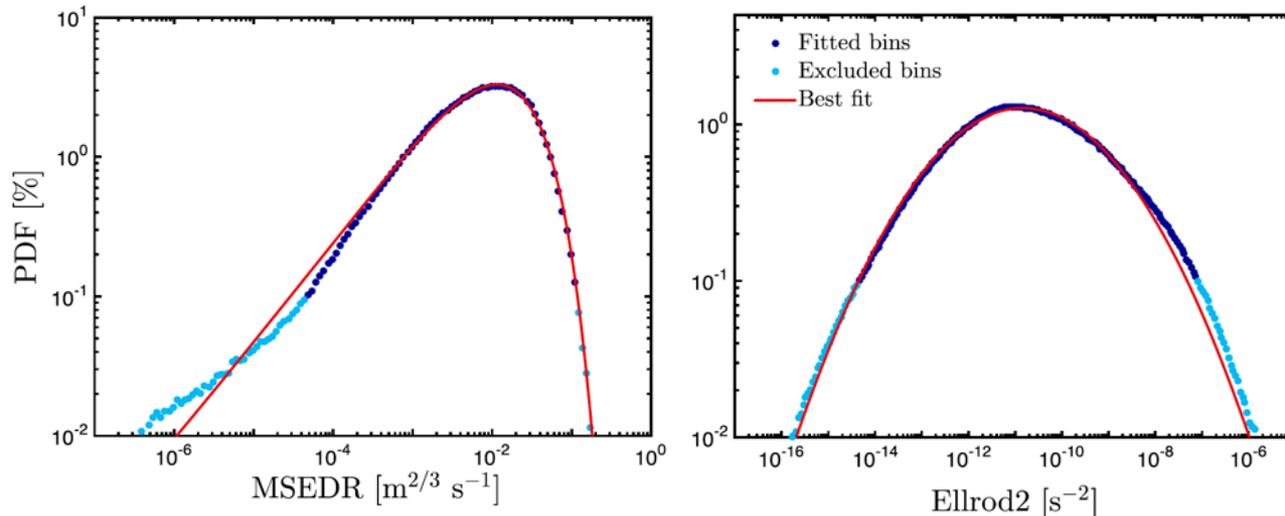
$$\boxed{\ln D^* = a + b \ln D}$$

$$a = \langle \ln \varepsilon^{1/3} \rangle - b \langle \ln D \rangle$$

$$b = \frac{SD[\ln \varepsilon^{1/3}]}{SD[\ln D]}$$

$$\left\{ \begin{array}{l} f(\ln x) = \frac{k}{\lambda} \left( \frac{\ln x}{\lambda} \right)^{k-1} e^{-(\ln x/\lambda)^k} \quad (\text{log - Weibull}) \\ f(\ln x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}} \quad (\text{lognormal}) \end{array} \right.$$

- 1-year long calibration using the High-Resolution Rapid Refresh (HRRR, 3 km)
- Optimization based on Mean Absolute Percentage Error (MAPE) minimization

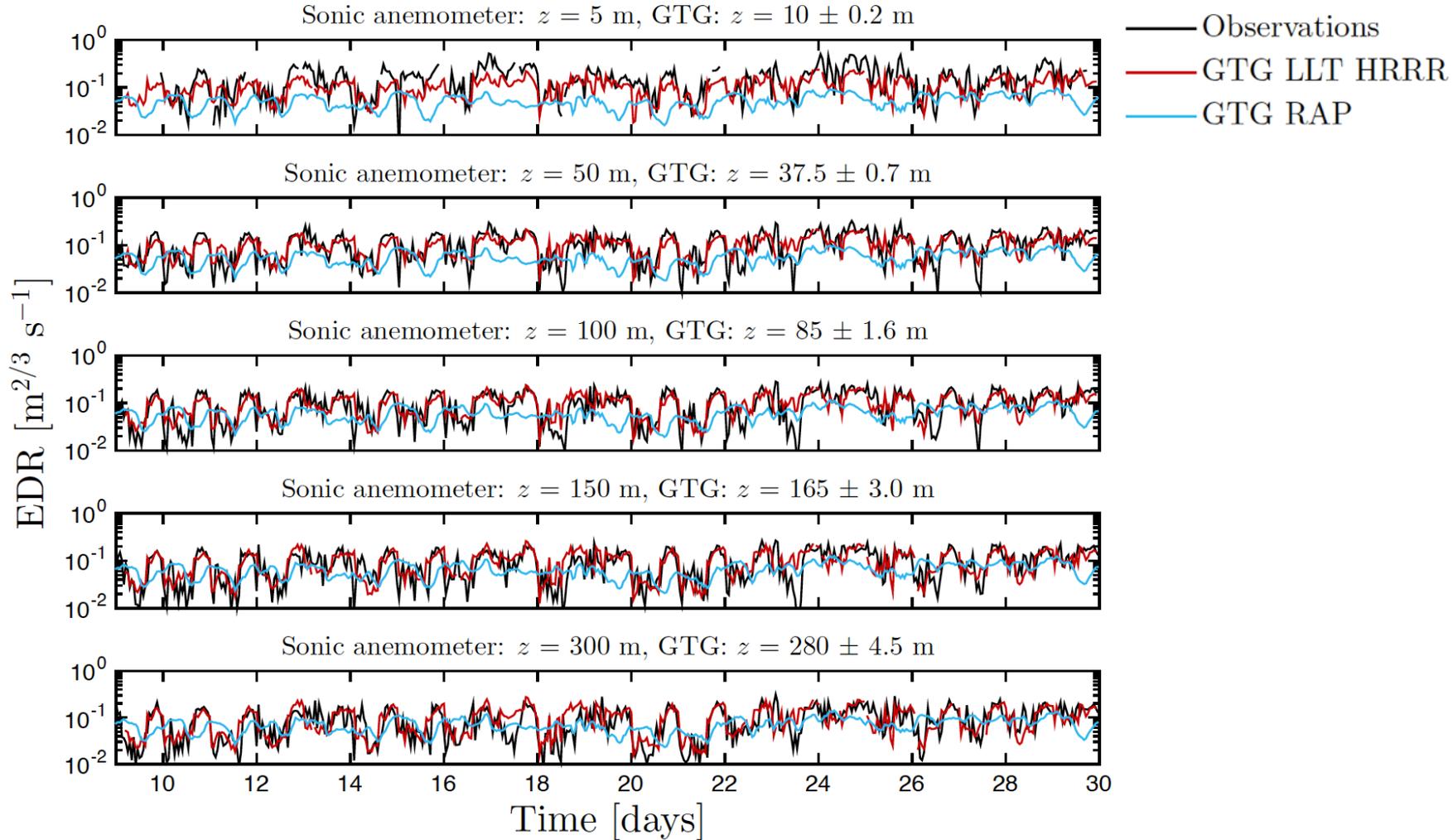


Muñoz-Esparza & Sharman (JAMC 2018)

# GTG LLT validation & comparison to GTG v3 (I)

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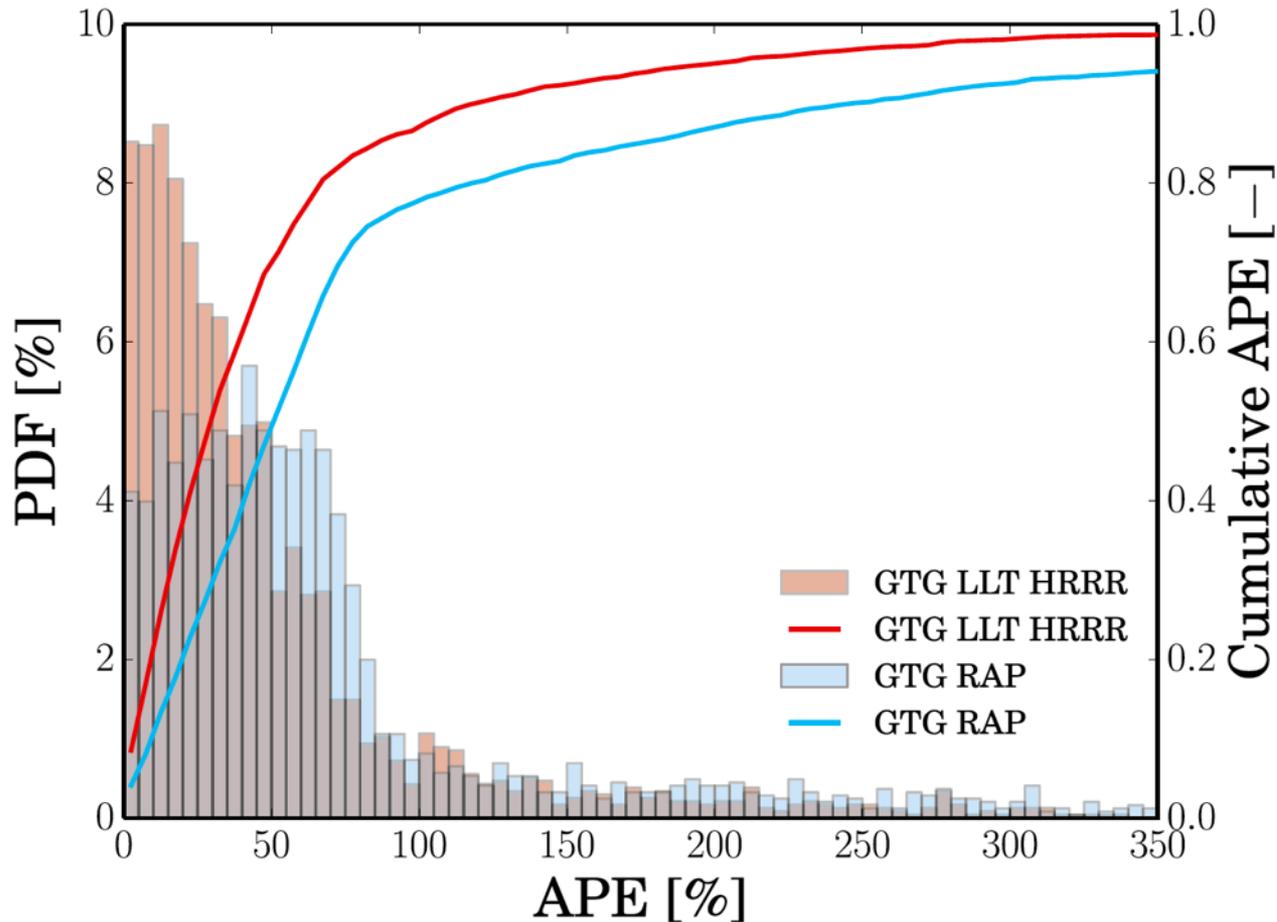
Improved ABL predictions based on specific “indices” and observation-derived understanding of climatological behavior of turbulence dissipation rate in the ABL



# GTG LLT validation & comparison to GTG v3 (II)

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CAT error decreased by a factor of 2 (MAPE = 55%) and ~20% increased probability of detection of typical low and high EDR values

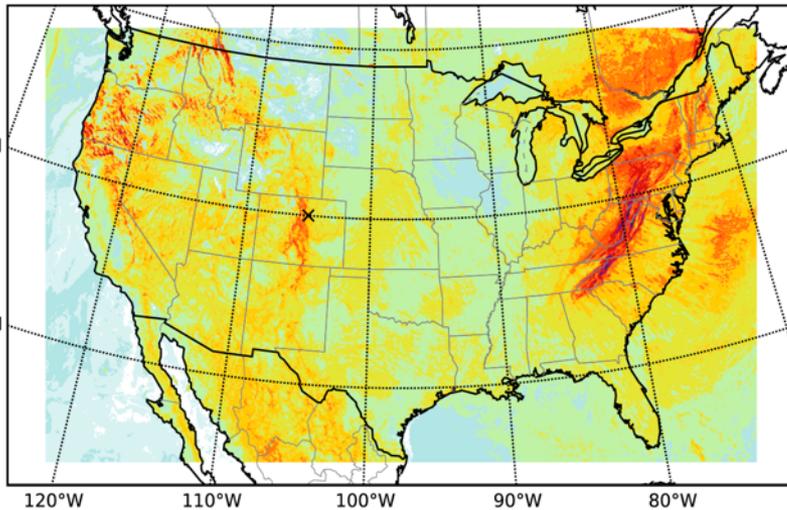


# GTG LLT validation & comparison to GTG v3

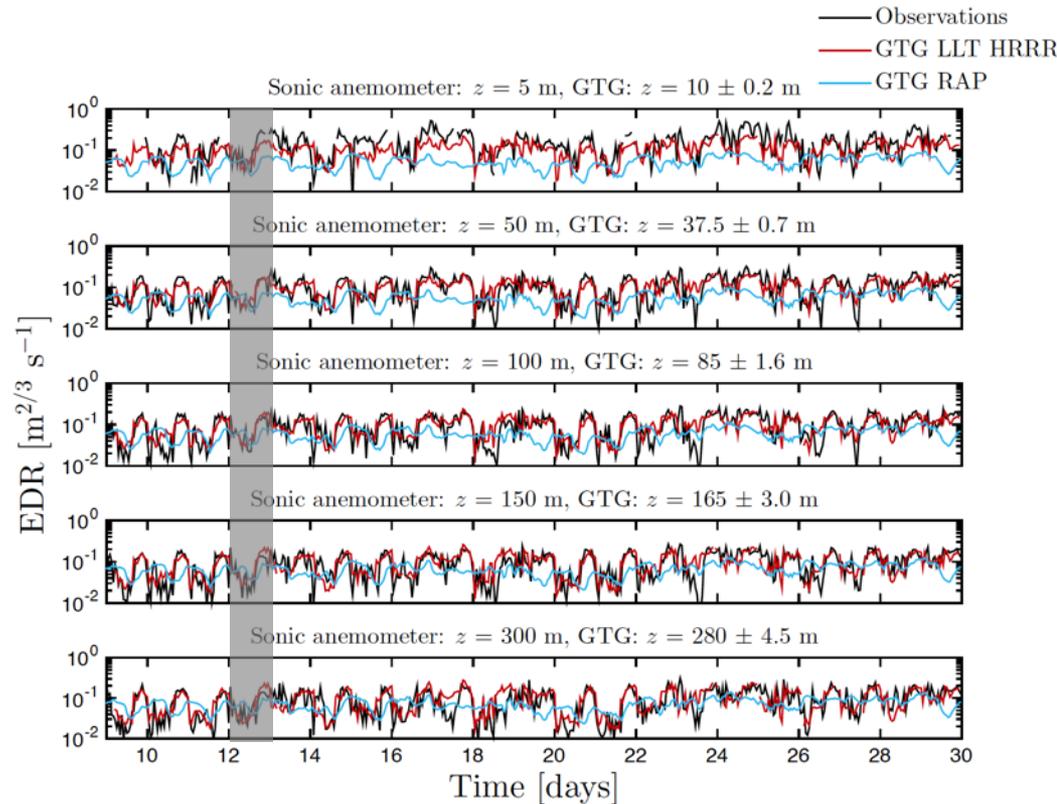
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- Improved ABL predictions based on specific “indices” and observation-derived understanding of climatological behavior of EDR in the ABL
- Implemented in operational G-GTG based on FV3 (code already delivered)
- GTG LLT will be part of GTG v4 (additionally supporting HRRR and RAP)

5h-fcst EDR,  $z = 150$  m (valid 18 UTC 15<sup>th</sup> March 2017)



X BAO tower



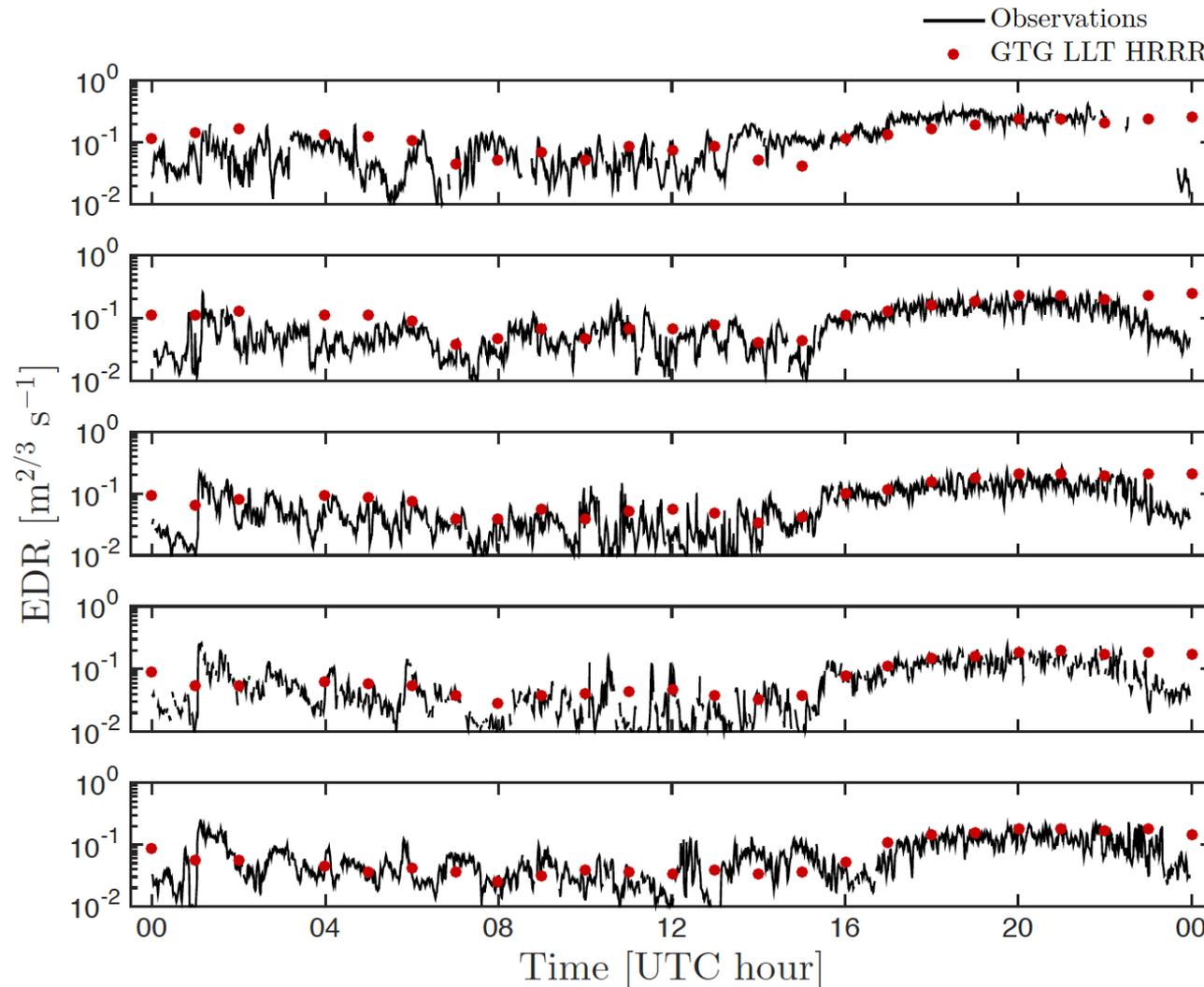
Overall diurnal evolution of ABL turbulence is correctly captured

# However...



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Significant variability is present at smaller spatiotemporal scales [s – min]



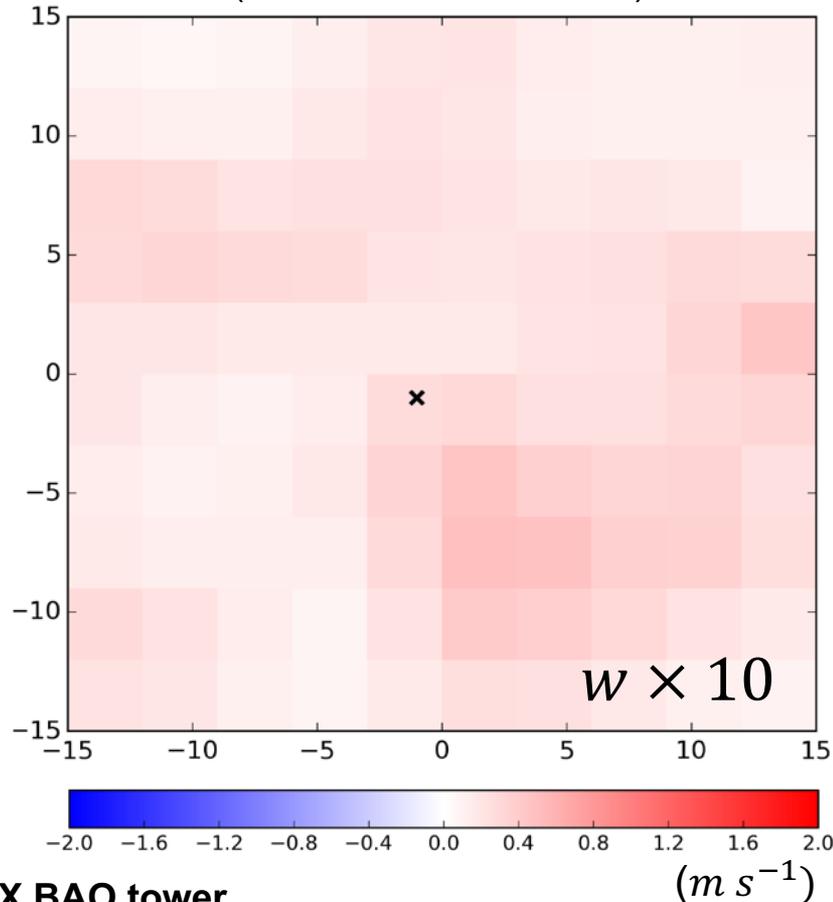
# Toward eddy-resolving forecasts



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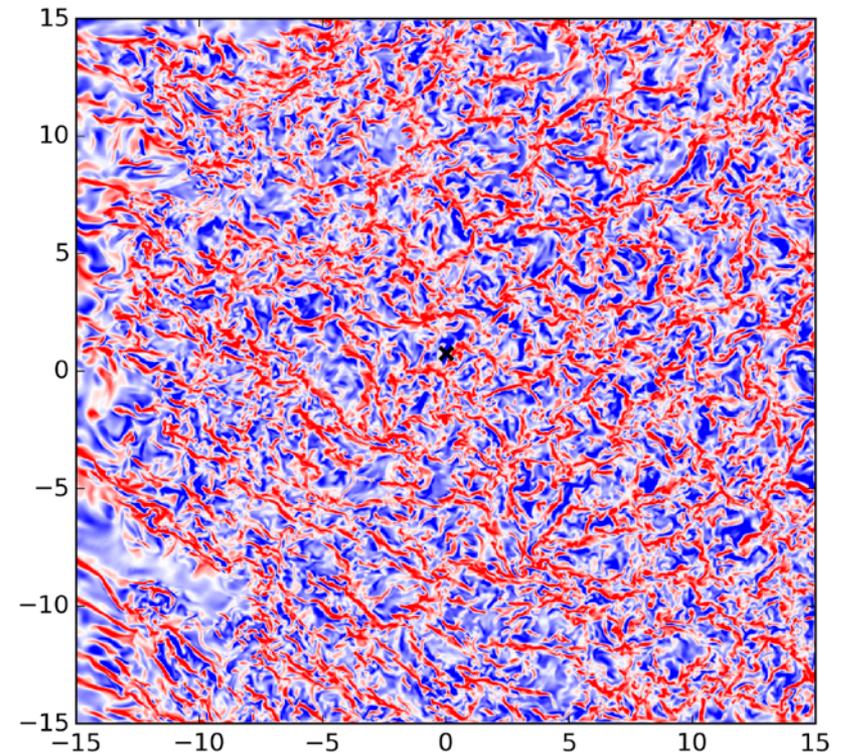
Operational NWP forecasts are too coarse to capture the required turbulent scales of interest...

5h-fcst vertical velocity,  $w$  [ $m\ s^{-1}$ ],  $z = 100\ m$   
(HRRR,  $\Delta = 3\ km$ ; 14 LT)



X BAO tower

5h-fcst vertical velocity,  $w$  [ $m\ s^{-1}$ ],  $z = 100\ m$   
(WRF mesoLES,  $\Delta = 25\ m$ ; 14 LT)



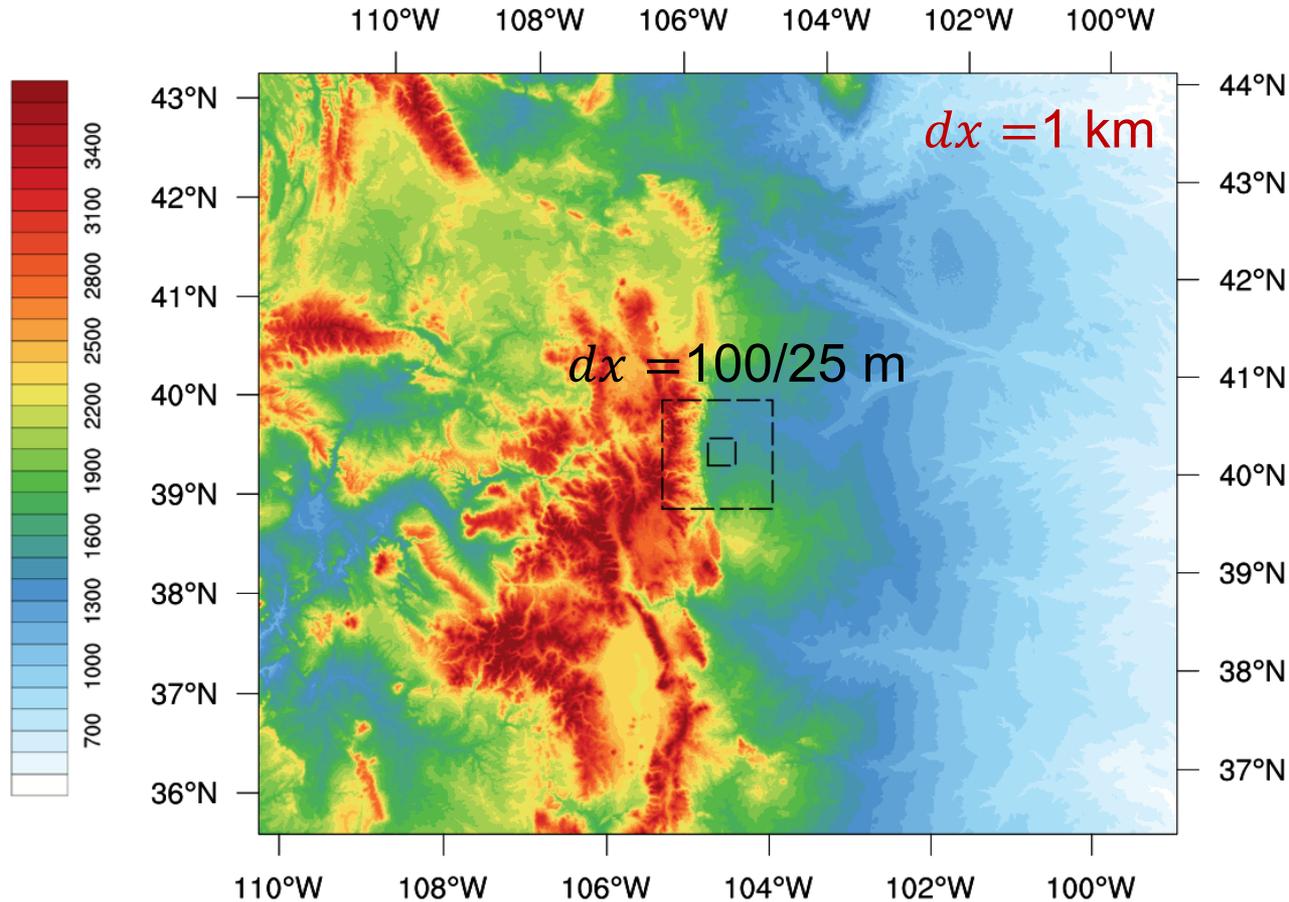
Muñoz-Esparza et al., GRL (2018)

# Toward eddy-resolving forecasts



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Dynamic downscaling from the RAP fcst (~13.2km) to nested 1 km mesoscale and 100m/25m LES [1200 x 1200 x 80 grid points]

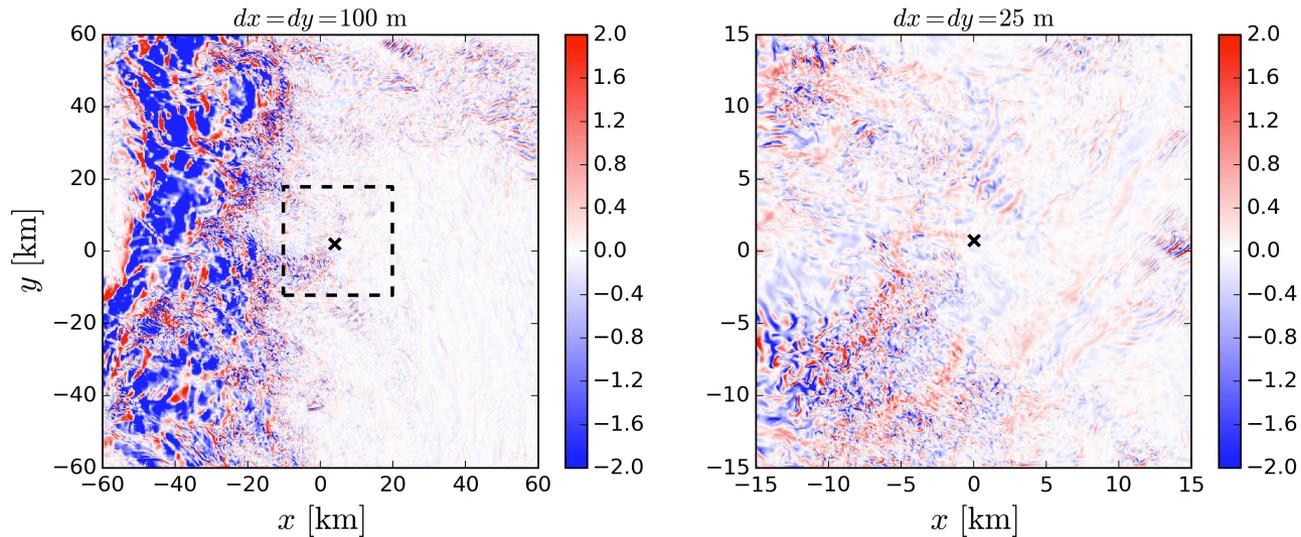


# Toward eddy-resolving forecasts

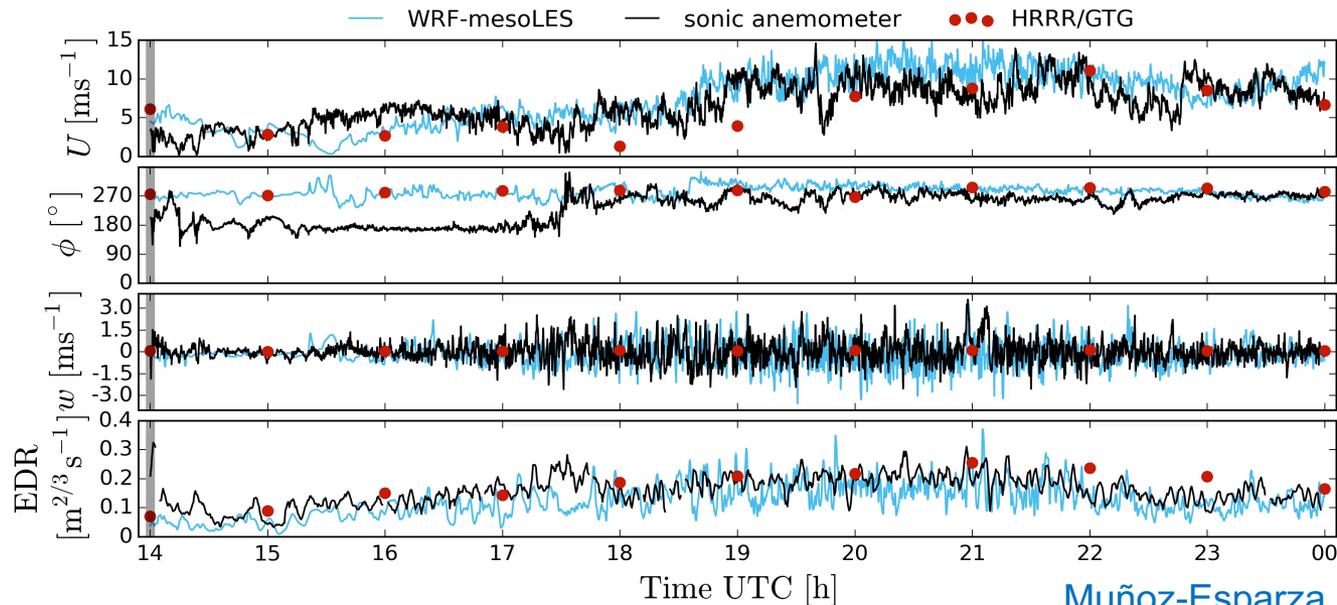


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Contours:  
vertical  
velocity  
[m s<sup>-1</sup>]  
(z = 100 m)



X BAO tower



Muñoz-Esparza et al., GRL (2018)

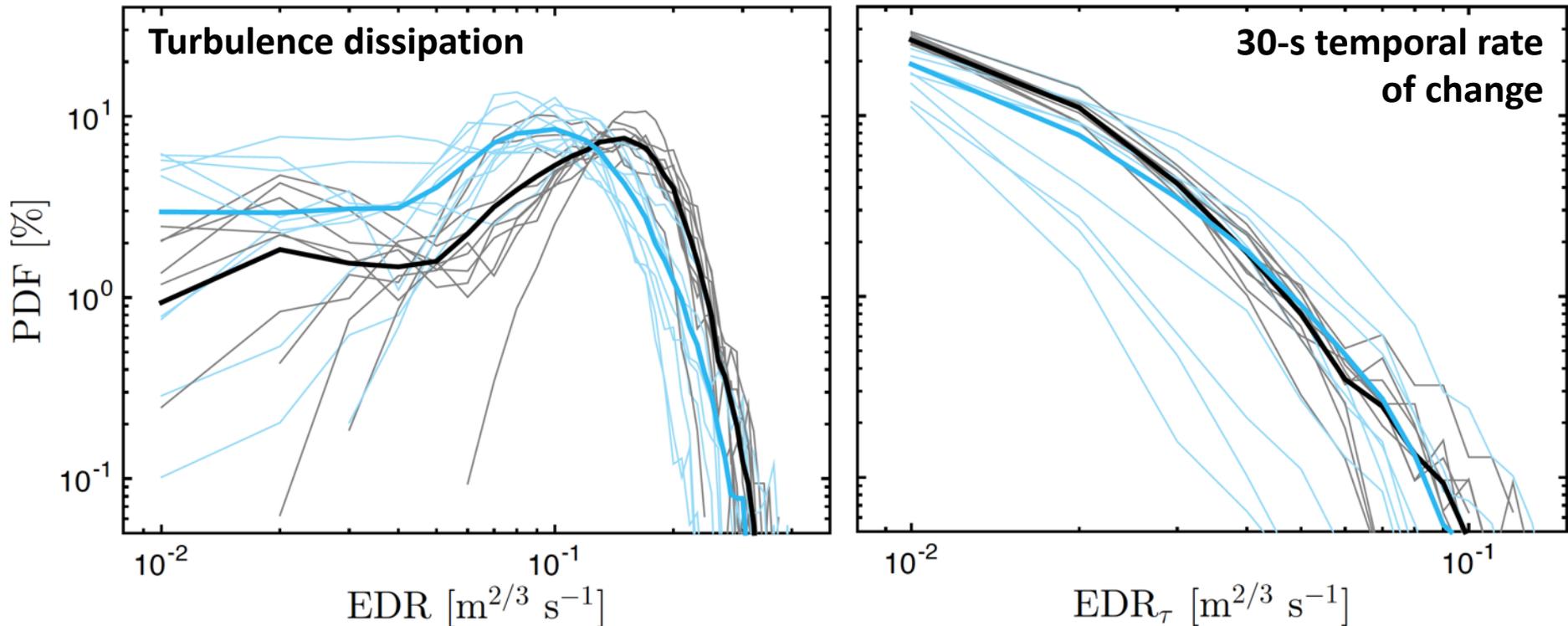
# Probabilistic turbulence forecasts



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Daytime portion (8am – 6 pm) of 10 days during the XPIA campaign with coupled WRF meso-LES [100 h total]

— Observations — Eddy-resolving forecasts

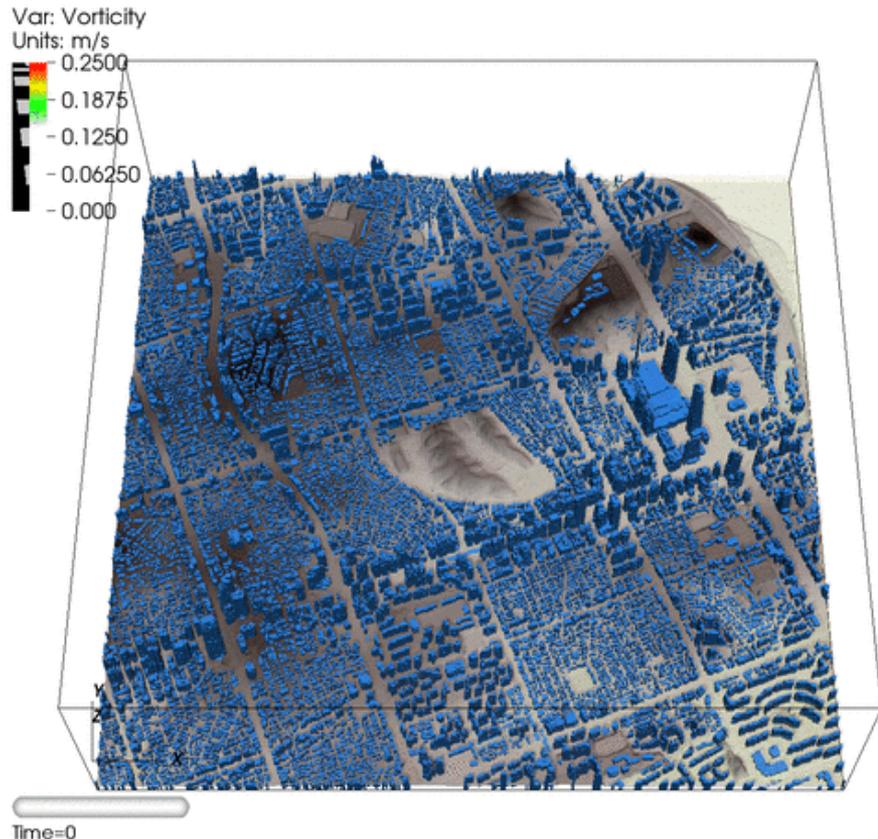


Both probability of EDR and its temporal rate of change are well reproduced by the eddy-resolving forecasts!

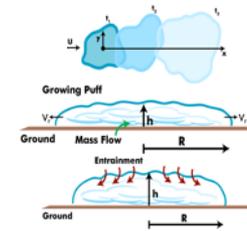
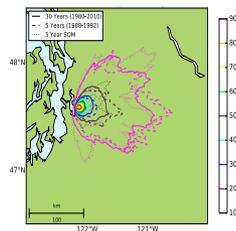
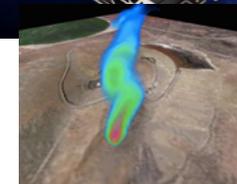
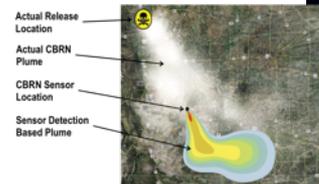
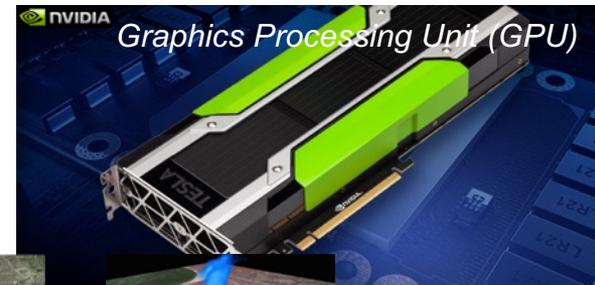
# Near-future directions: urban environment

Significant interest on urban scenarios

- High resolutions and building-resolving capabilities are required
- Accelerated GPU-LES to enable real-time forecasting applications



NCAR-RAL team is developing an accelerated GPU-LES code: “FastEddy”



Lee & Muñoz-Esparza  
(JAMC 2018, submitted)

# Conclusions



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- ✓ We have enhanced GTG with the development of a LLT-specific algorithm → accurately represents diurnal evolution of EDR in the ABL
- ✓ GTG LLT forecasting in the ABL has the potential to benefit: UAVs, wake turbulence ...
- ✓ Meter-scale coupled WRF mesoscale-LES realistically reproduces turbulence in the ABL [dissipation rate validated through comparison to sonic anemometer data] → need to move toward “probabilistic” LES-scale forecasts
- ✓ NCAR’s GPU-accelerated LES model FastEddy will enable a path toward high-resolution eddy-resolving forecasts in the near term [including building-resolving urban capabilities]

# Thanks for your attention!!!

## “Improvements in Low Level Turbulence (LLT) forecasting for UAVs”

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