

UCL-UM Ur

- 1. ExUM and Radar data trends
- 2. Tidal analysis
- 3. Next steps



1. Data Ir

Data ExUM Radar

Data ExUM Radar

- 1D-CIRA
- 3D-CIRA
- DTM

Data Radar

• 1D-CIRA

EXUM

- 3D-CIRA
- DTM

- Ascension
- Esrange
- Rothera

Data

EXUM

3-hourly

- 1D-CIRA
- 3D-CIRA
- DTM

Radar

hourly

- Ascension
- Esrange
- Rothera

Data

EXUM

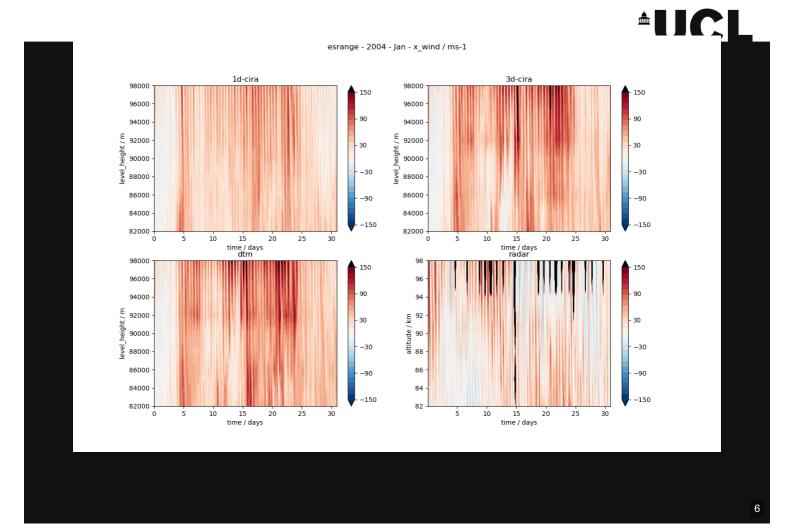
3-hourly

- 1D-CIRA
- 3D-CIRA
- DTM

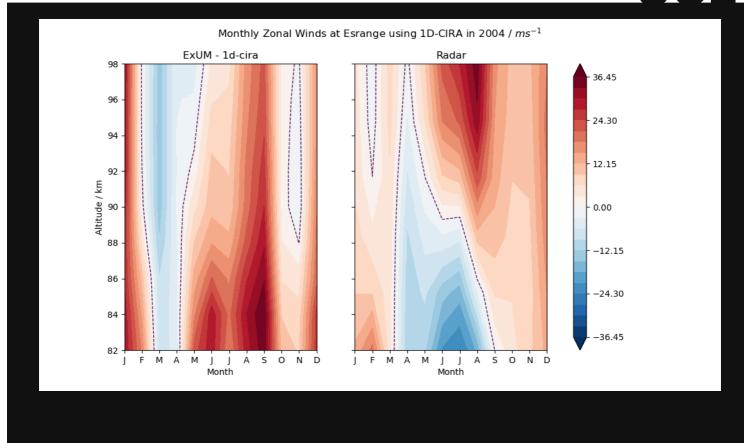
Radar

hourly

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2. Tidal Ana



Composite Days

Following the work of *Matthew J. Griffith et al. (2021)* we create a sequence of 4-day composite days.



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Defined at discrete altitudes in 82-98km.



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Defined at discrete altitudes in 82-98km.

For radar data **only**.

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Model

$$egin{aligned} y_i &= A_0 + A_1 \sin(2t_i\pi/24 + \omega_1) \ &+ A_2 \sin(2t_i\pi/12 + \omega_2) \ &+ A_3 \sin(2t_i\pi/8 + \omega_3) \ &+ A_4 \sin(2t_i\pi/6 + \omega_4) + \epsilon_i \end{aligned}$$

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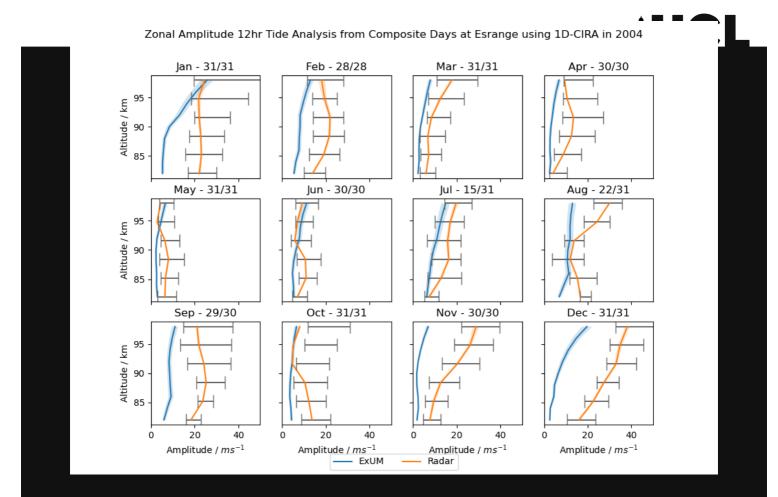
Model

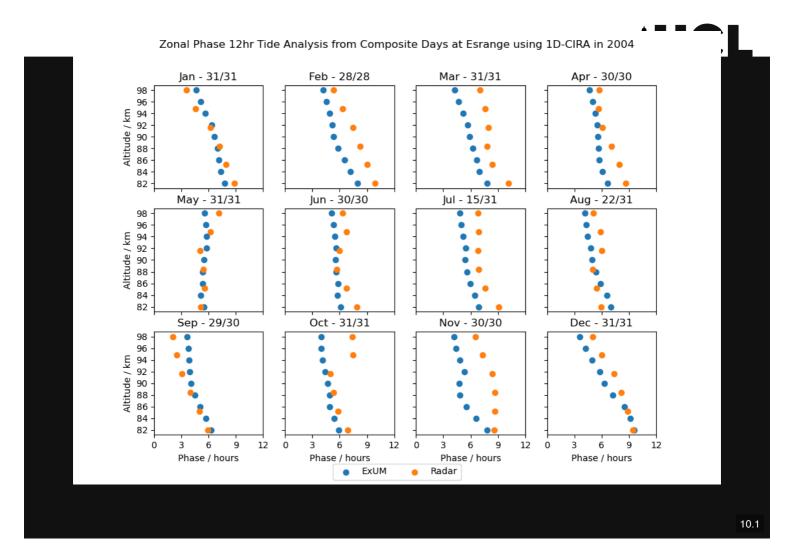
$$egin{aligned} y_i &= A_0 + A_1 \sin(2t_i\pi/24 + \omega_1) \ &+ A_2 \sin(2t_i\pi/12 + \omega_2) \ &+ A_3 \sin(2t_i\pi/8 + \omega_3) \ &+ A_4 \sin(2t_i\pi/6 + \omega_4) + \epsilon_i \end{aligned}$$

ExUM

Radar

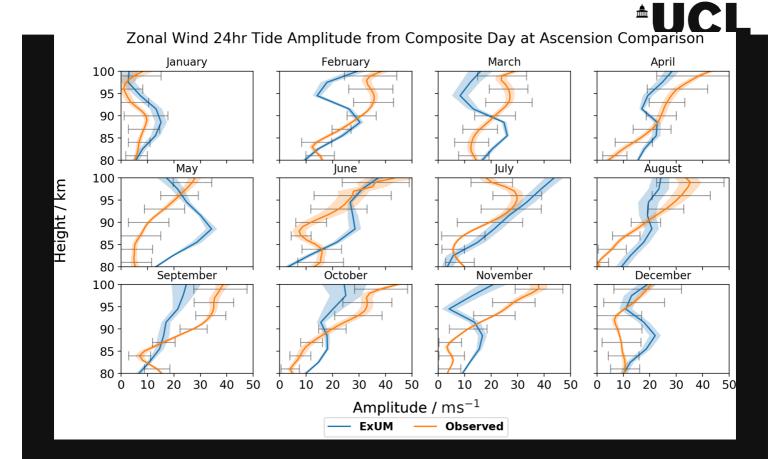
4-day Composites







"The shaded regions denote the standard deviation from the curve-fitting algorithm, and the black bars indicate the standard deviation from the mean of the measured amplitudes across the month."



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3. Next S



More Data

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More Data

 More plots requires more simulation data from other years. We have the radar data!



More Data

- More plots requires more simulation data from other years. We have the radar data!
- Improved regression parameter estimates needs more UM output points. eg hourly output.

1D-CIRA

$$T_{nudged} = lpha_{NP} T_{profile} + (1-lpha_{NP}) T_{dynamic}$$
 where $lpha_{NP} = rac{k\Delta t}{ au}$

1D-CIRA

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 where $lpha_{NP} = rac{k\Delta t}{ au}$

- au chosen to be 24hours in keeping with Song et al. (2018)
- ullet $k \in [0,1]$ is a 'nudging ramp parameter'.

1D-CIRA

 $\overline{T_{profile}}$ is based on climatological and satellite data:

- 70-86 km is based on the US Standard Atmosphere (USSA) (COESA, 1976).
- 86-119.7 km is based on the Committee on Space Research (COSPAR) Intern
- 119.7+ km the temperature asymptotes to a selected exobase temperature

```
T_{\text{PROFILE}}(z) \ = \left\{ \begin{array}{l} T_{70 \text{ km}} - \Gamma_{\text{MESO}} \left( z - z_{\text{USSA-bottom}} \right) & \text{, if } 70 \leq z \leq 86 \text{ km} \\ T_{70 \text{ km}} - \Gamma_{\text{MESO}} (z_{\text{USSA-top}} - z_{\text{USSA-bottom}}) & \text{, if } 86 < z \leq 93.3 \text{ km} \\ T_{70 \text{ km}} - \Gamma_{\text{MESO}} (z_{\text{USSA-top}} - z_{\text{USSA-bottom}}) & \text{, if } 86 < z \leq 93.3 \text{ km} \\ -\Gamma_{\text{CIRA}} \left( z - z_{\text{USSA-top}} \right) & \text{, if } 93.3 < z \leq 119.7 \text{ km} \\ + (3 \times 10^{-7}) (z - z_{\text{Tmin-CIRA}})^2 & \text{, if } 93.3 < z \leq 119.7 \text{ km} \\ T_{\text{exobase}} - \left( T_{\text{exobase}} - T_{\text{CIRA-top}} \right) \exp \left( -\Gamma_{\text{THERMO}} \sigma(z) \right) & \text{, if } z > 119.7 \text{ km} \end{array} \right.
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3D-CIRA

$$egin{aligned} T_m(m,arphi) &= T' + T_a\cos(|m-6|\pi/6)\cos(\pi/2+arphi) \ z_m(m,arphi) &= z' + z_a\cos(|m-6|\pi/6)\cos(\pi/2+arphi) \end{aligned}$$

3D-CIRA

$$T_m(m,arphi) = T' + T_a\cos(|m-6|\pi/6)\cos(\pi/2 + z_m(m,arphi)) = z' + z_a\cos(|m-6|\pi/6)\cos(\pi/2 + arphi)$$

$$m$$
 = month, $arphi$ = latitude $T'=178.4482K$, $T_a=25.73031K$ $z'=94.066km$, $z_a=4.561km$

3D-CIRA

$$T(m,arphi,z)=T_m(m,arphi)\Gamma(z-z_m(m,arphi))^a$$

3D-CIRA

$$T(m,arphi,z)=T_m(m,arphi)\Gamma(z-z_m(m,arphi))^a \ \Gamma=4.025643 imes10^{-6}Kkm^{-1} \ a=2.414574$$



DTM

This model is much more complex. Split into 2 parts: 120-170km and 170-200km

It involves polynomial (only quadratic?) fitting to the amplitudes from DTM output.

Eddy Diffusion

 K_{zz} drives eddy diffusion and is defined analytically.

It is inferred from SABER atomic oxygen measurements using a similar method to the tidal analysis seen earlier.

[RD-25] says K_{zz} varies with the solar cycle.

