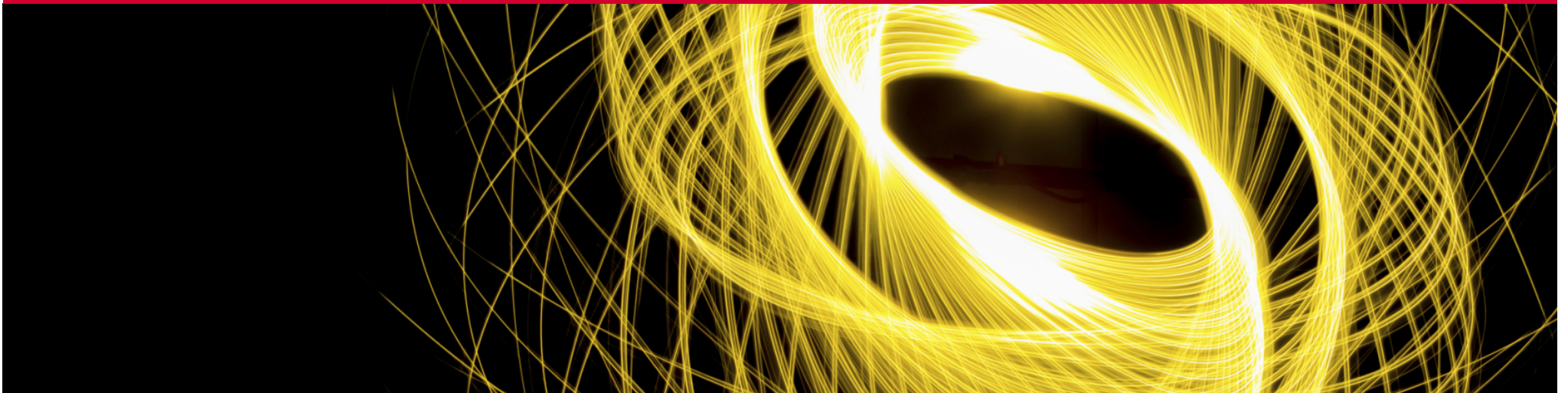


# Statistical *vs/and* Physical prediction



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*With particular thanks to students, postdocs and collaborators in the Energy-Meteorology research group*



**National Centre for  
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# Weather and climate

- Drivers: differential insolation and rotation
- Atmospheric heat engine: transparent to SW/visible, opaque to long wave/IR
- Large scale structure of atmospheric/ocean circulation relatively well understood
- ... but great complexity for simulation and prediction

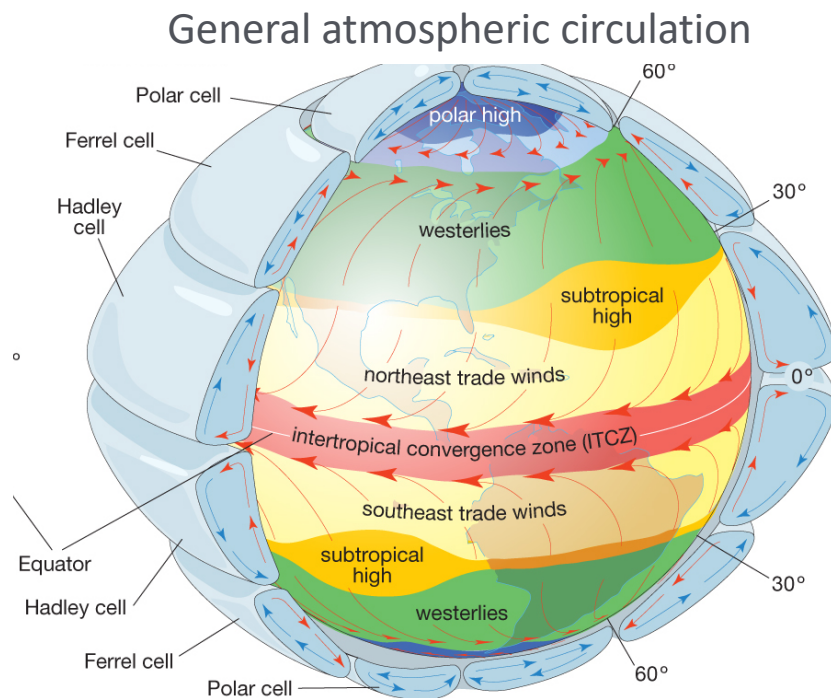
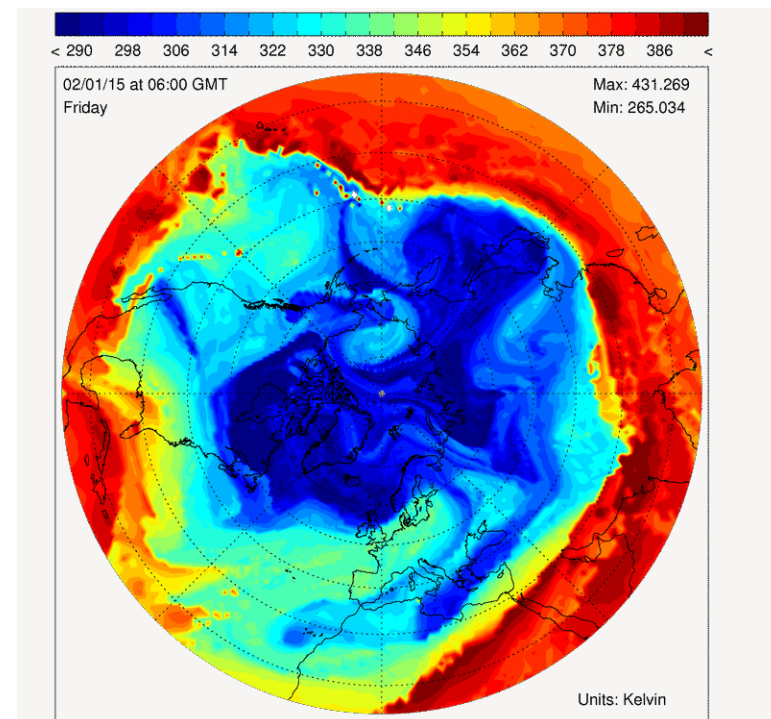


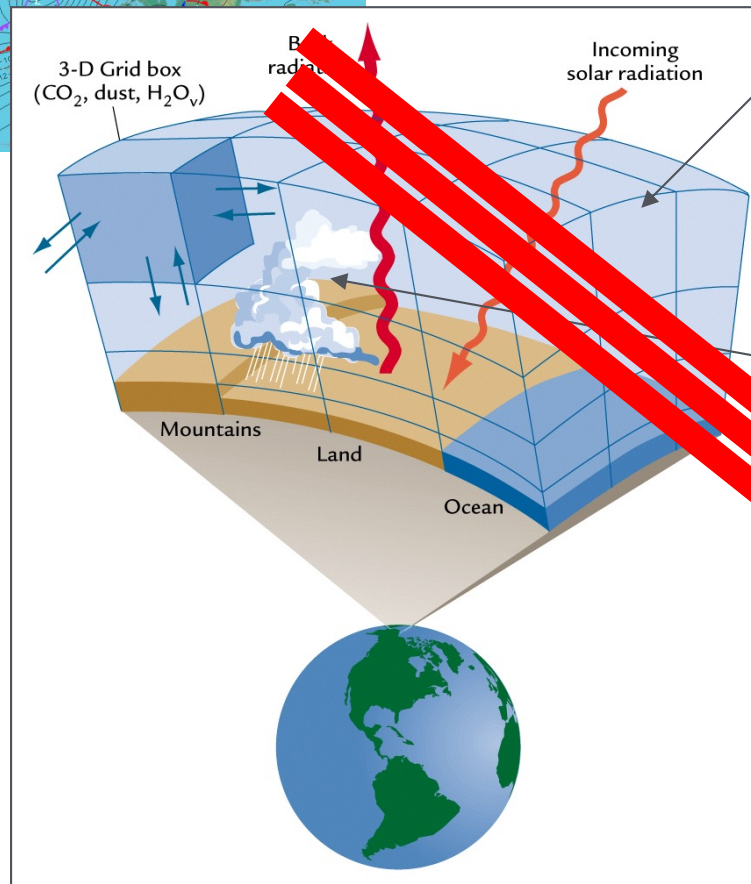
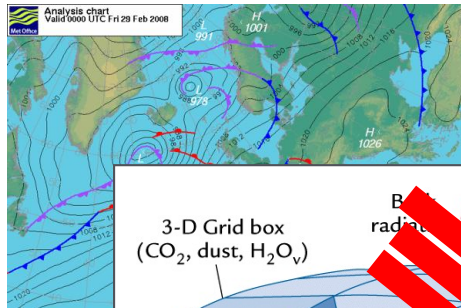
Fig: Encyclopedia Britanica

Theta on PV2  
(summarises atmospheric circulation)



# Numerical Weather Prediction

## Initial state analysis



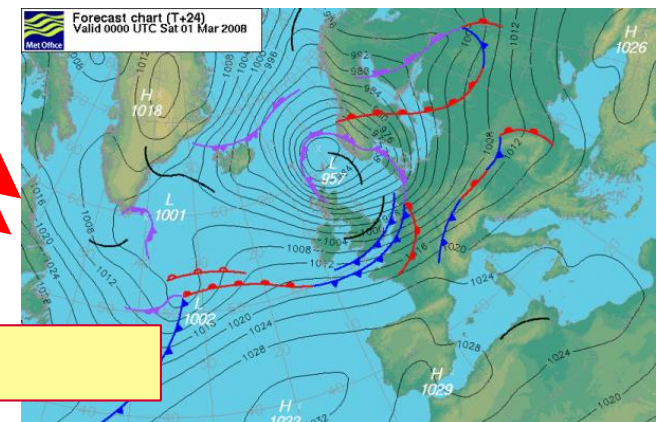
## Dynamical “core” ~ based on Navier-Stokes eqns

$$\rho \left( \frac{\partial u_r}{\partial t} + u_r \frac{\partial u_r}{\partial r} + \frac{u_\theta}{r \sin(\phi)} \frac{\partial u_r}{\partial \theta} + \frac{u_\phi}{r} \frac{\partial u_r}{\partial \phi} - \frac{u_\theta^2 + u_\phi^2}{r} \right) = -\frac{\partial p}{\partial r} + \rho g_r$$
$$\mu \left[ \frac{1}{r^2} \frac{\partial}{\partial r} \left( r^2 \frac{\partial u_r}{\partial r} \right) + \frac{1}{r^2 \sin(\phi)^2} \frac{\partial^2 u_r}{\partial \theta^2} + \frac{1}{r^2 \sin(\phi)} \frac{\partial}{\partial \phi} \left( \sin(\phi) \frac{\partial u_r}{\partial \phi} \right) - 2 \frac{u_r + \frac{\partial u_\phi}{\partial \phi} + u_\phi \cot(\phi)}{r^2} + \frac{2}{r^2 \sin(\phi)} \frac{\partial u_\theta}{\partial \theta} \right]$$
$$\rho \left( \frac{\partial u_\theta}{\partial t} + u_r \frac{\partial u_\theta}{\partial r} + \frac{u_\theta}{r \sin(\phi)} \frac{\partial u_\theta}{\partial \theta} + \frac{u_\phi}{r} \frac{\partial u_\theta}{\partial \phi} + \frac{u_r u_\theta + u_\theta u_\phi \cot(\phi)}{r} \right) = -\frac{1}{r \sin(\phi)} \frac{\partial p}{\partial \theta} + \rho g_\theta$$
$$\mu \left[ \frac{1}{r^2} \frac{\partial}{\partial r} \left( r^2 \frac{\partial u_\theta}{\partial r} \right) + \frac{1}{r^2 \sin(\phi)^2} \frac{\partial^2 u_\theta}{\partial \theta^2} + \frac{1}{r^2 \sin(\phi)} \frac{\partial}{\partial \phi} \left( \sin(\phi) \frac{\partial u_\theta}{\partial \phi} \right) + \frac{2 \frac{\partial u_r}{\partial r} + 2 \cos(\phi) \frac{\partial u_\phi}{\partial \phi} - u_\theta}{r^2 \sin(\phi)^2} \right]$$
$$\rho \left( \frac{\partial u_\phi}{\partial t} + u_r \frac{\partial u_\phi}{\partial r} + \frac{u_\theta}{r \sin(\phi)} \frac{\partial u_\phi}{\partial \theta} + \frac{u_\phi}{r} \frac{\partial u_\phi}{\partial \phi} + \frac{u_r u_\phi - u_\theta^2 \cot(\phi)}{r} \right) = -\frac{1}{r} \frac{\partial p}{\partial \phi} + \rho g_\phi$$
$$\mu \left[ \frac{1}{r^2} \frac{\partial}{\partial r} \left( r^2 \frac{\partial u_\phi}{\partial r} \right) + \frac{1}{r^2 \sin(\phi)^2} \frac{\partial^2 u_\phi}{\partial \theta^2} + \frac{1}{r^2 \sin(\phi)} \frac{\partial}{\partial \phi} \left( \sin(\phi) \frac{\partial u_\phi}{\partial \phi} \right) + \frac{2}{r^2} \frac{\partial u_r}{\partial r} - \frac{u_\phi + 2 \cos(\phi) \frac{\partial u_\theta}{\partial \theta}}{r^2 \sin(\phi)^2} \right]$$

## Physical “parameterisation” schemes

- Cloud/moisture processes
- Surface energy and momentum exchange
- Radiation...

Forecast  
(ensemble)

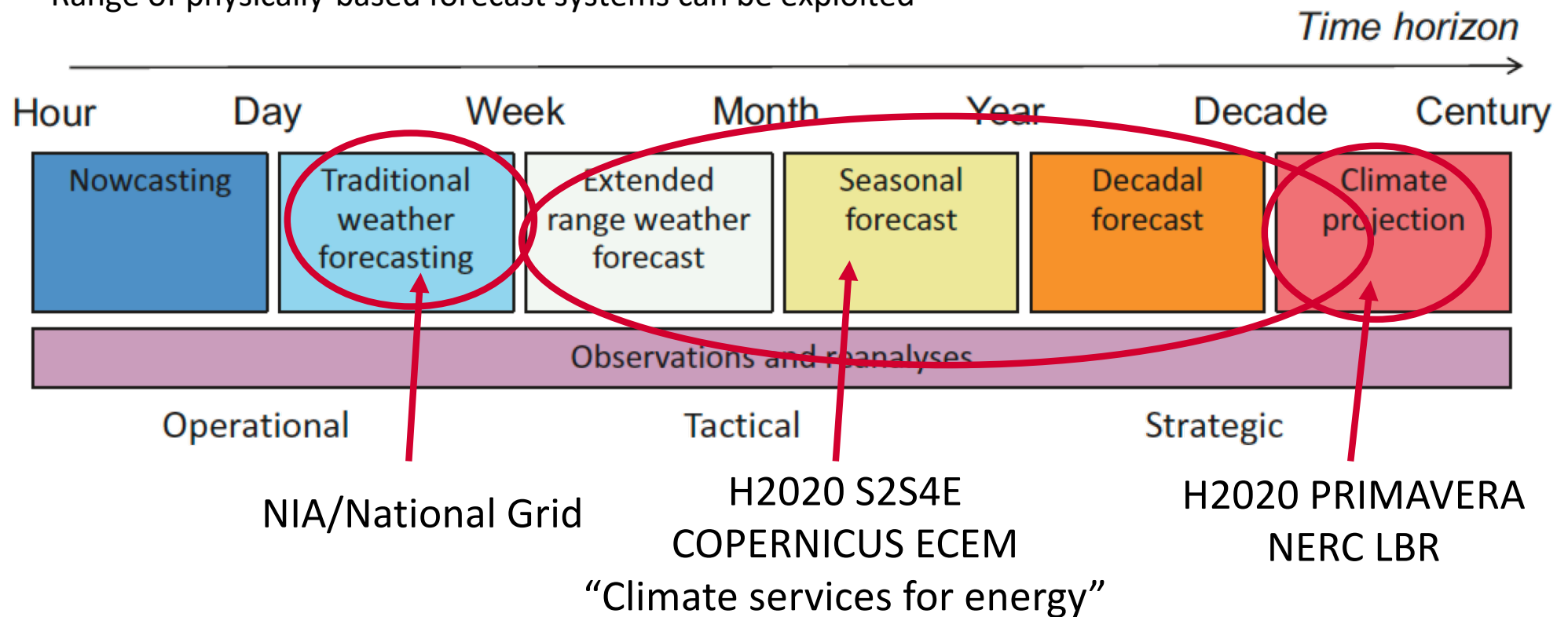


Expensive simulations, large supercomputers, large datasets



# Introduction

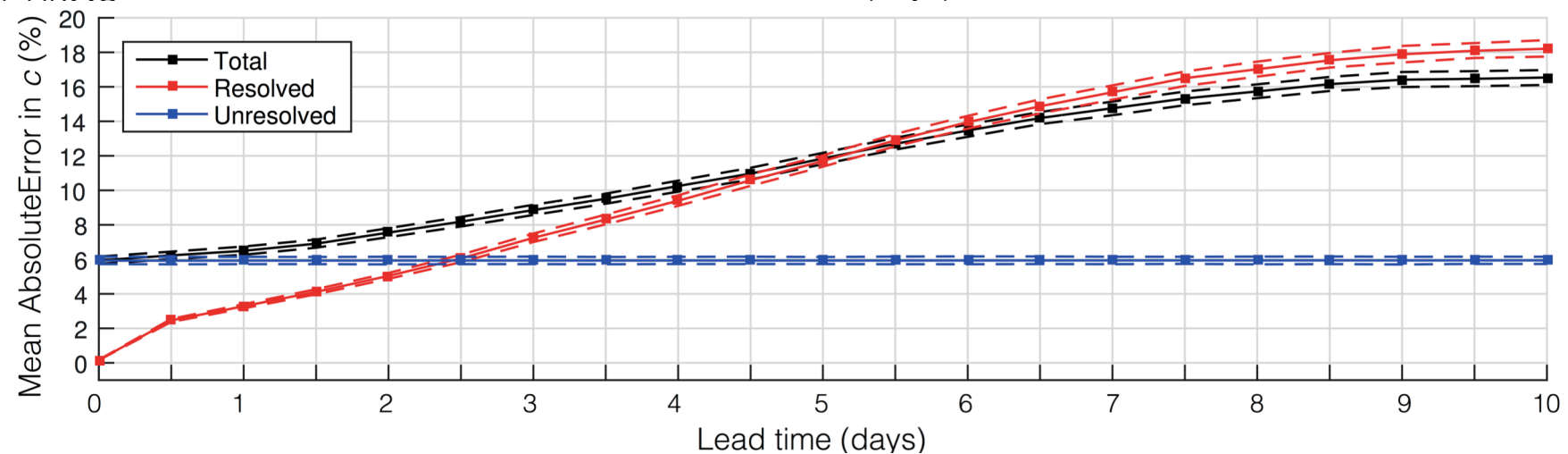
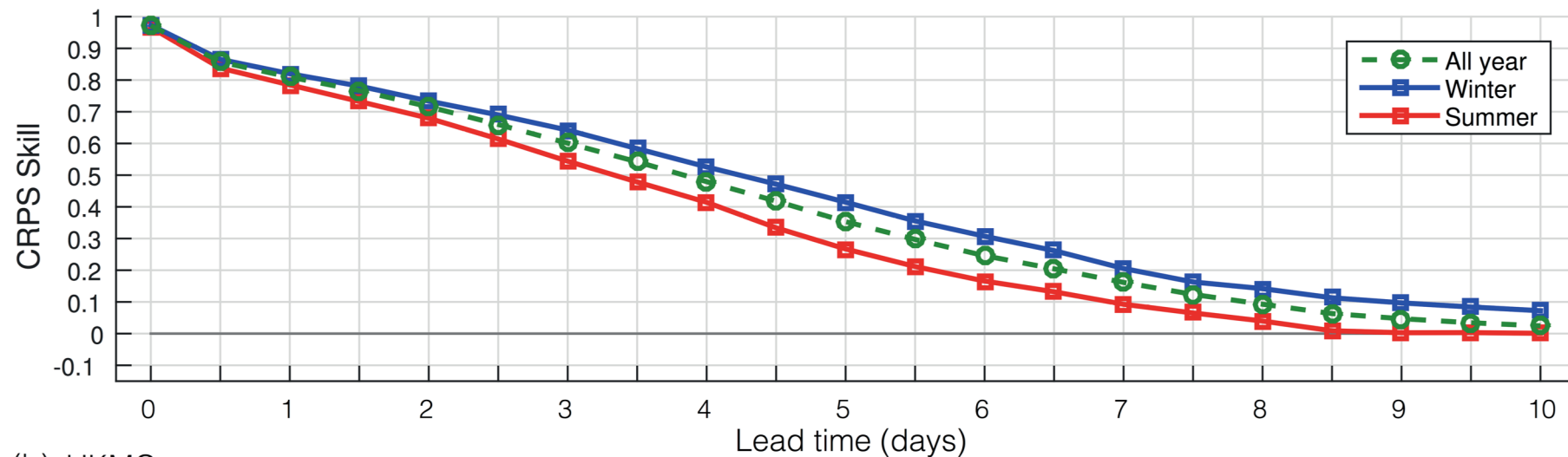
- Climate is more than just the atmosphere – variability and predictability on many timescales
- Power system operations is more than just real-time / day-ahead balancing, e.g.:
  - Trading
  - Maintenance scheduling
- Range of physically-based forecast systems can be exploited



# Example 1: days-ahead forecasting

## *Cannon et al, 2017 (Met Zet)*

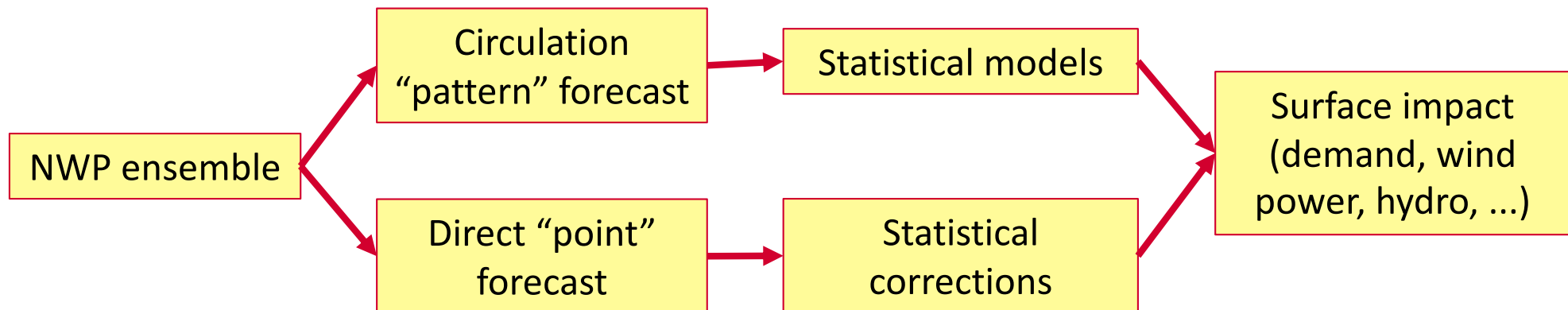
- Instantaneous GB-total wind power generation: probabilistic skill to ~9 days (ECMWF)
- Decomposition: resolved (large-scale, upper limit), unresolved (large-scale to site, lower limit)
  - Statistical methods unlikely to could greatly improve UR but unlikely to improve R



# Example 2: weeks-months ahead

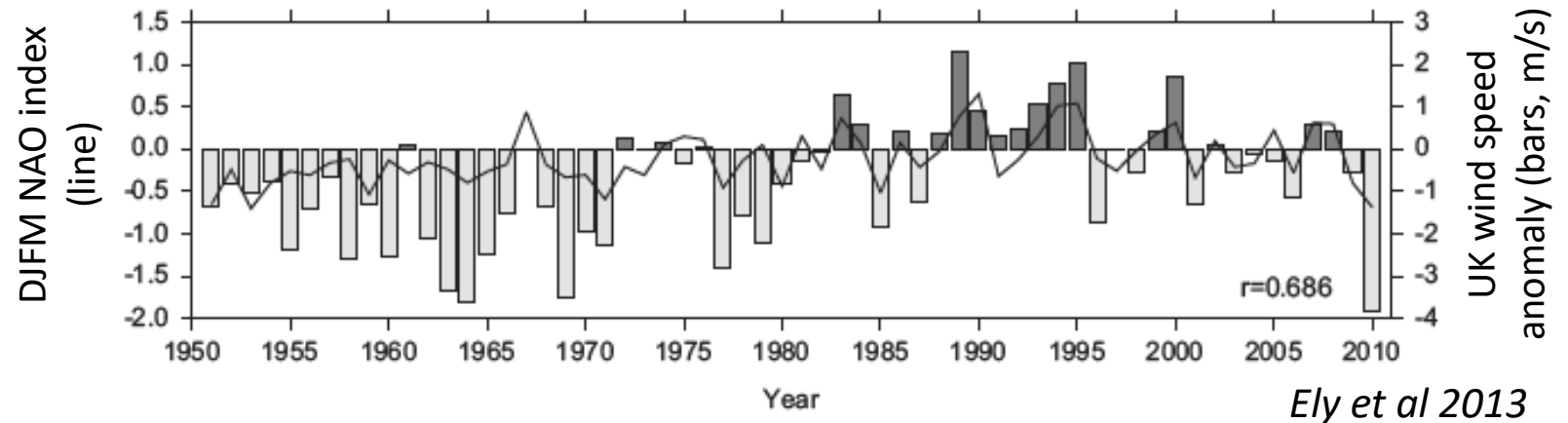
## *The growth of climate services*

- Skill in forecasting weekly-/monthly-/seasonal- demand, wind, ...
- Energy is a major target sector for climate service development
  - ECEM and CLIM4POWER (EU Copernicus); e.g.: <http://ecem.wemcouncil.org>
  - S2S4E, SECLIFIRM, CLIM4RES (EU H2020); e.g.: <https://s2s4e.eu>
  - Ongoing interest/activities at National, International and Private Weather Services



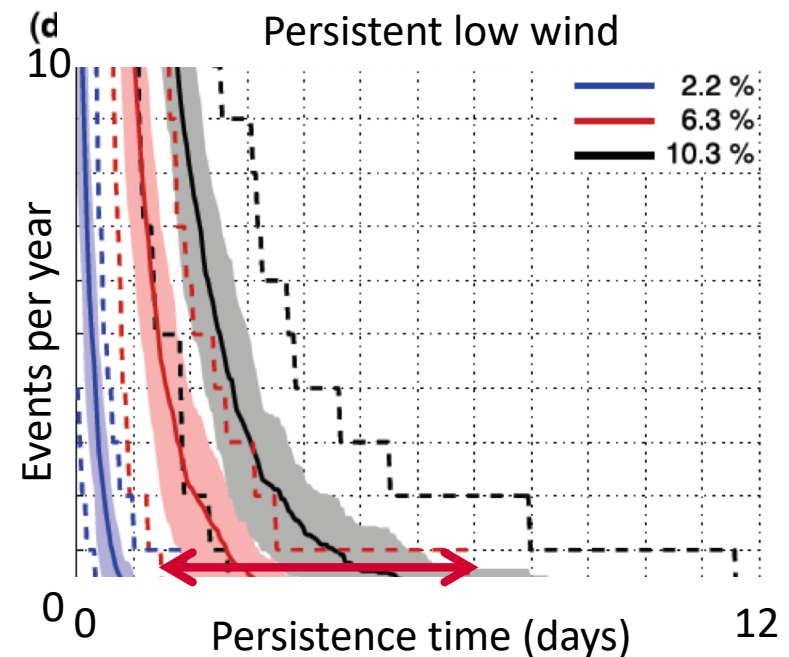
- Some specific issues:
  - Pattern choices
  - Statistical models and corrections

# Example 3: Zeroth kind of predictability?



- Most naive forecast: climatological PDF
- Beware: Non-stationarity across timescales *even if anthropogenic climate change neglected*

Red line → CF<6%  
Solid lines = mean value  
Shading = 1 stddev  
Dotted lines = extreme range



Cannon et al 2015

# Challenges and opportunities

- Physically-based forecasts can be a powerful tool days-seasons ahead
- Complementarity: statistical AND physical
  - Combination techniques
  - Bias correction
  - Downscaling / calibration
  - Hybrid approaches – e.g., AnEn, pattern forecasting (e.g., Bett et al (in prep); Bloomfield et al (in prep))

## General:

- Opportunities around climate service development
- Evaluation issues:
  - Understanding decision-making
  - Multi-variate trajectory forecasts



# References and contact details

Contact: [d.j.brayshaw@reading.ac.uk](mailto:d.j.brayshaw@reading.ac.uk)

Website: <https://research.reading.ac.uk/met-energy/>

Major projects:

- ECEM climate service demonstrator: <http://ecem.climate.copernicus.eu>
- S2S4E (Subseasonal-to-seasonal forecasting for energy): please email for details
- PRIMAVERA advanced climate modelling: <https://uip.primavera-h2020.eu>

Selected papers:

- Brayshaw (2018) Weather, climate and the nature of predictability. In Weather and Climate Services for the Energy Industry (ed. A. Troccoli)
- Brayshaw (2018) The nature of weather and climate impacts in the energy sector. In Weather and Climate Services for the Energy Industry (ed. A. Troccoli)
- Cannon, D. et al. (2017) Determining the bounds of skilful forecast range for probabilistic prediction of system-wide wind power generation. Meteorologische Zeitschrift, 26 (3). pp. 239-252.
- Cannon, D.J. et al (2015) Using reanalysis data to quantify extreme wind power generation statistics : a 33 year case study in Great Britain. Renewable Energy, 75. pp. 767-778.
- Thornton, H. E. et al (2017) The relationship between wind power, electricity demand and winter weather patterns in Great Britain. Environmental Research Letters, 12 (6). 064017.