

The Gauging and Modelling of Rivers in the Sky

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EUROPEAN CENTRE FOR MEDIUM-RANGE WEATHER
FORECASTS

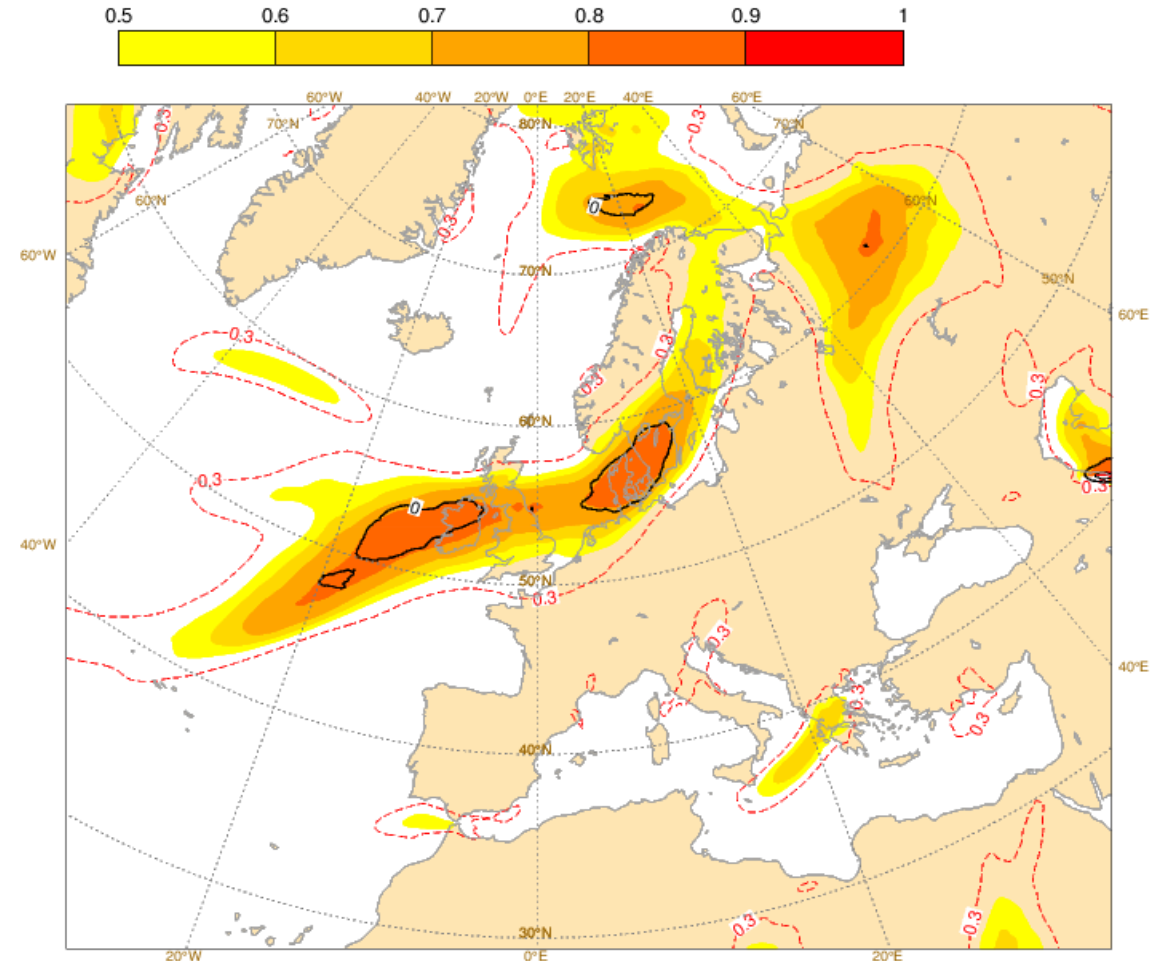


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Atmospheric River Forecasting

- Extreme Forecast Index (EFI) for water vapour flux.
- Compares ensemble forecast with the model climate.
- May provide increased awareness of extreme rainfall and ARs.
- Implemented in 46r1.

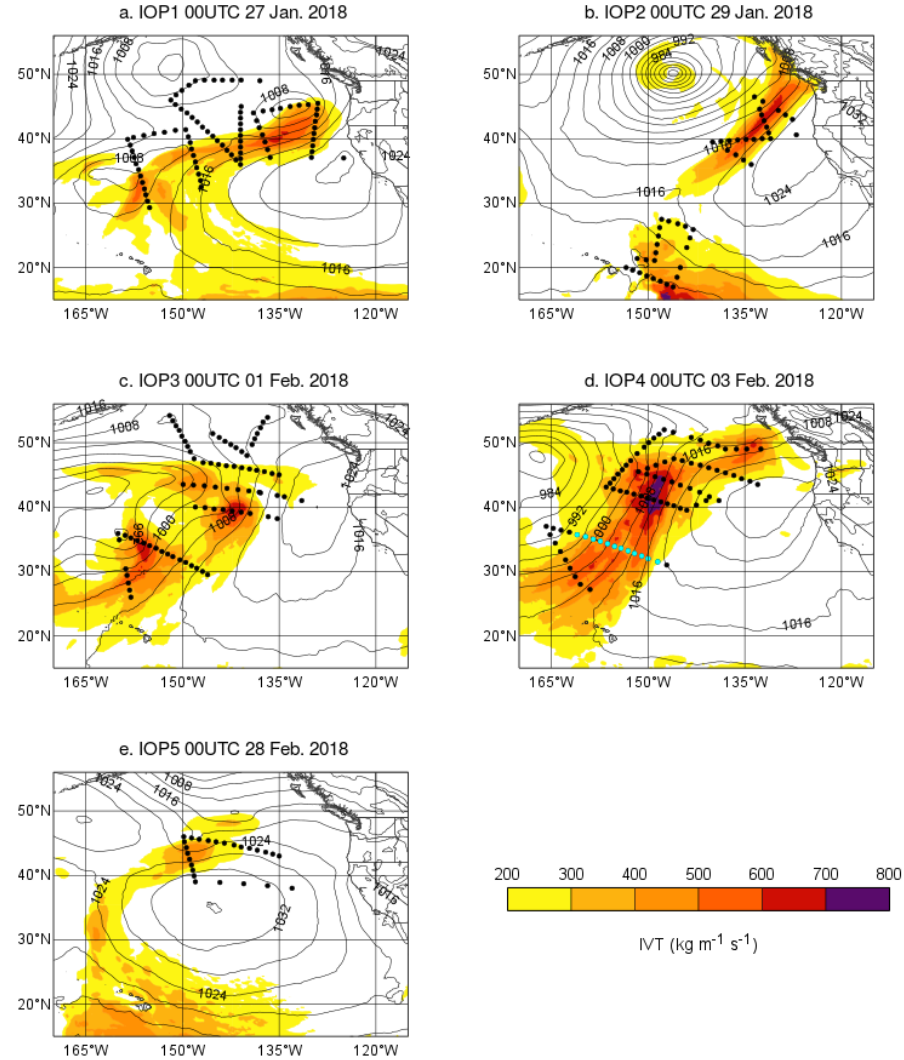
Tue 28 May 2019 00UTC ©ECMWF t+048-072h VT: Thu 30 May 2019 00UTC - Fri 31 May 2019 00UTC
Extreme forecast index and Shift of Tails (black contours 0,1,2,5,8) for water vapour flux



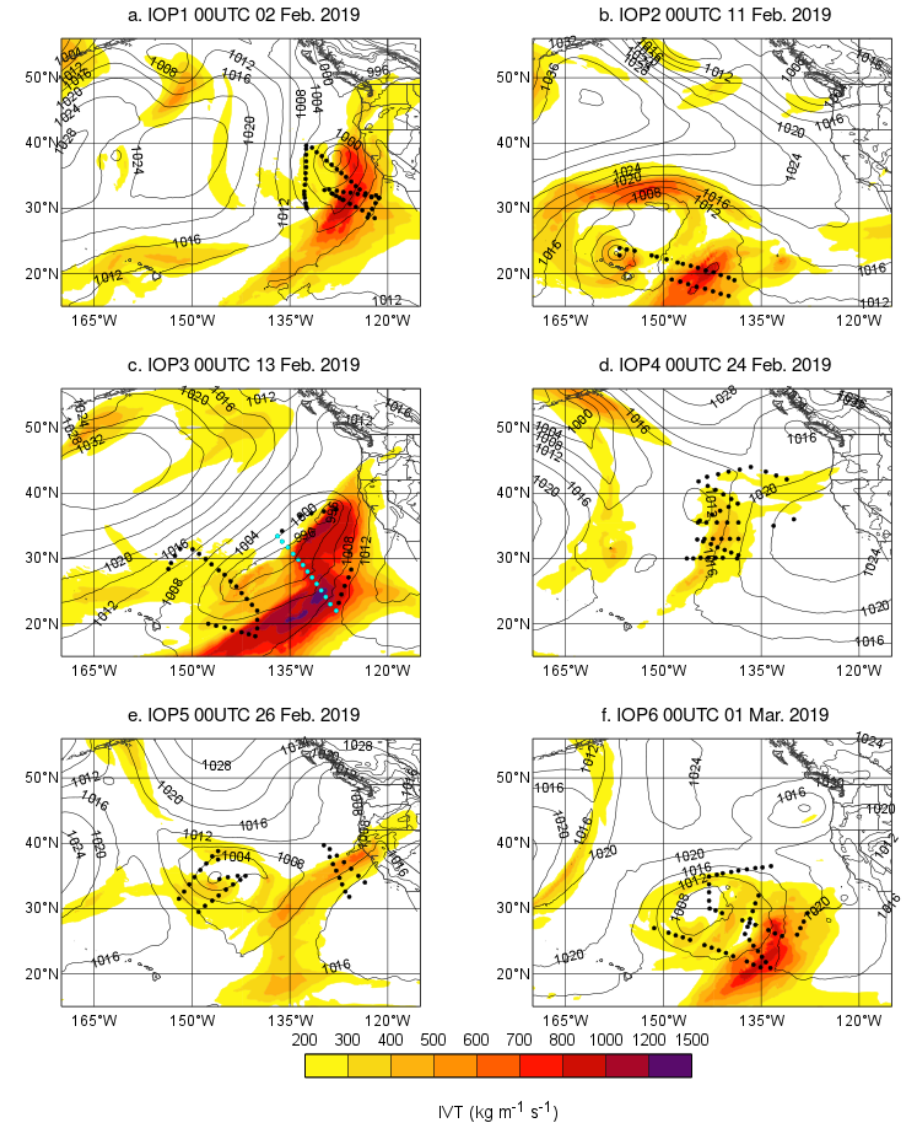
Atmospheric River Reconnaissance (AR recon)

- January / February 2018 and February / March 2019.
- 12 Intensive Observation Periods (IOPs).
- Three aircraft (NOAA GIV and two C130s).
- ARs are important for extreme rainfall and atmospheric circulation and predictability.
- Opportunity to identify model problems.

2018



2019



Analysis – part 1

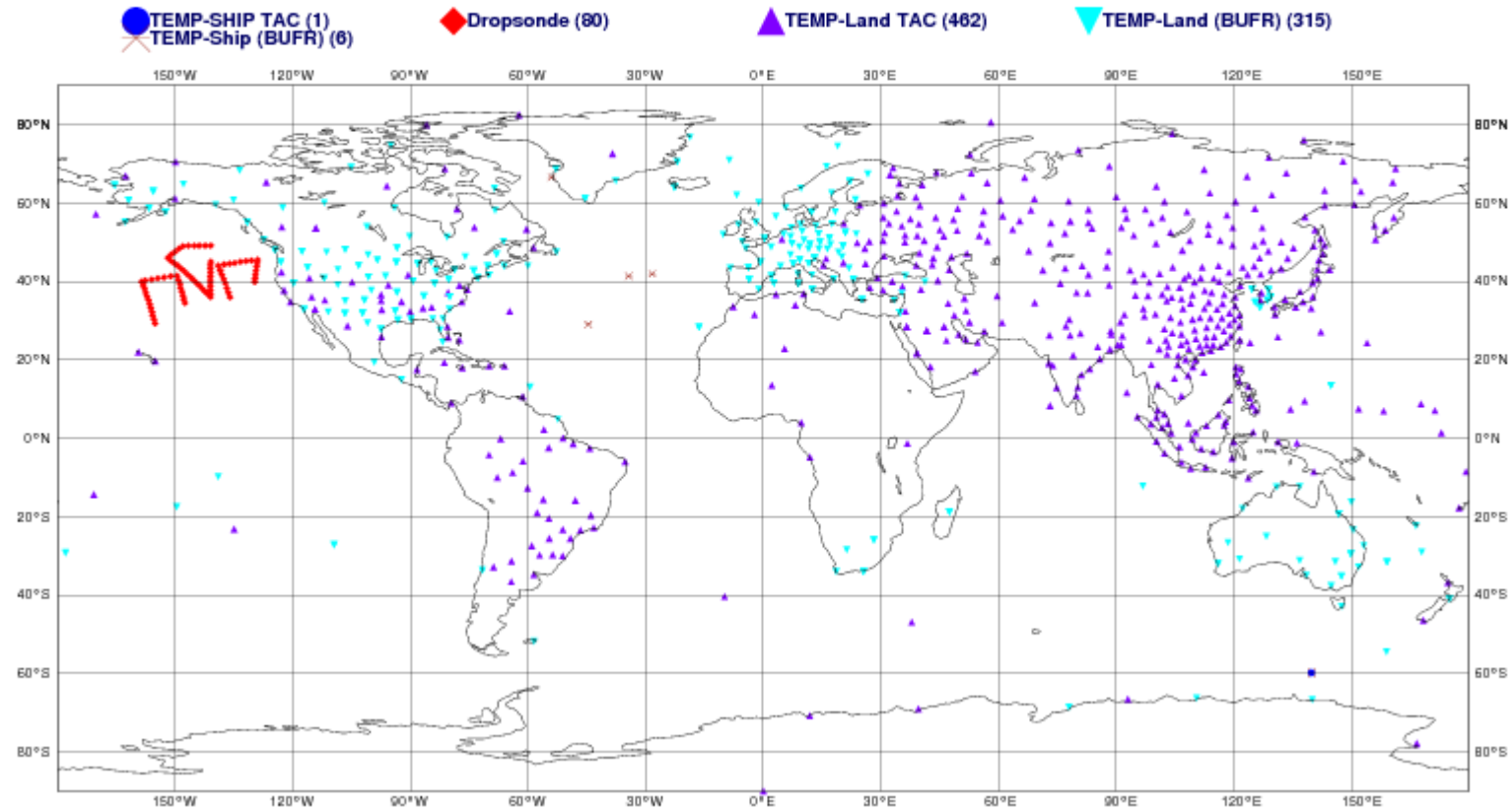
- Dropsondes were assimilated in real-time.
- Use ECMWF ensemble of data assimilations (EDA). At the time, the 25 members produced the 50 perturbed ensemble forecasts.
- Water vapour flux (IVT) calculated at each dropsonde location.
- Mostly used three pressure levels (925, 850, 700 hPa).
- Assess the background, analysis, and observed values.
- AR transects evaluated and IVT uncertainties investigated.

Assimilated dropsondes

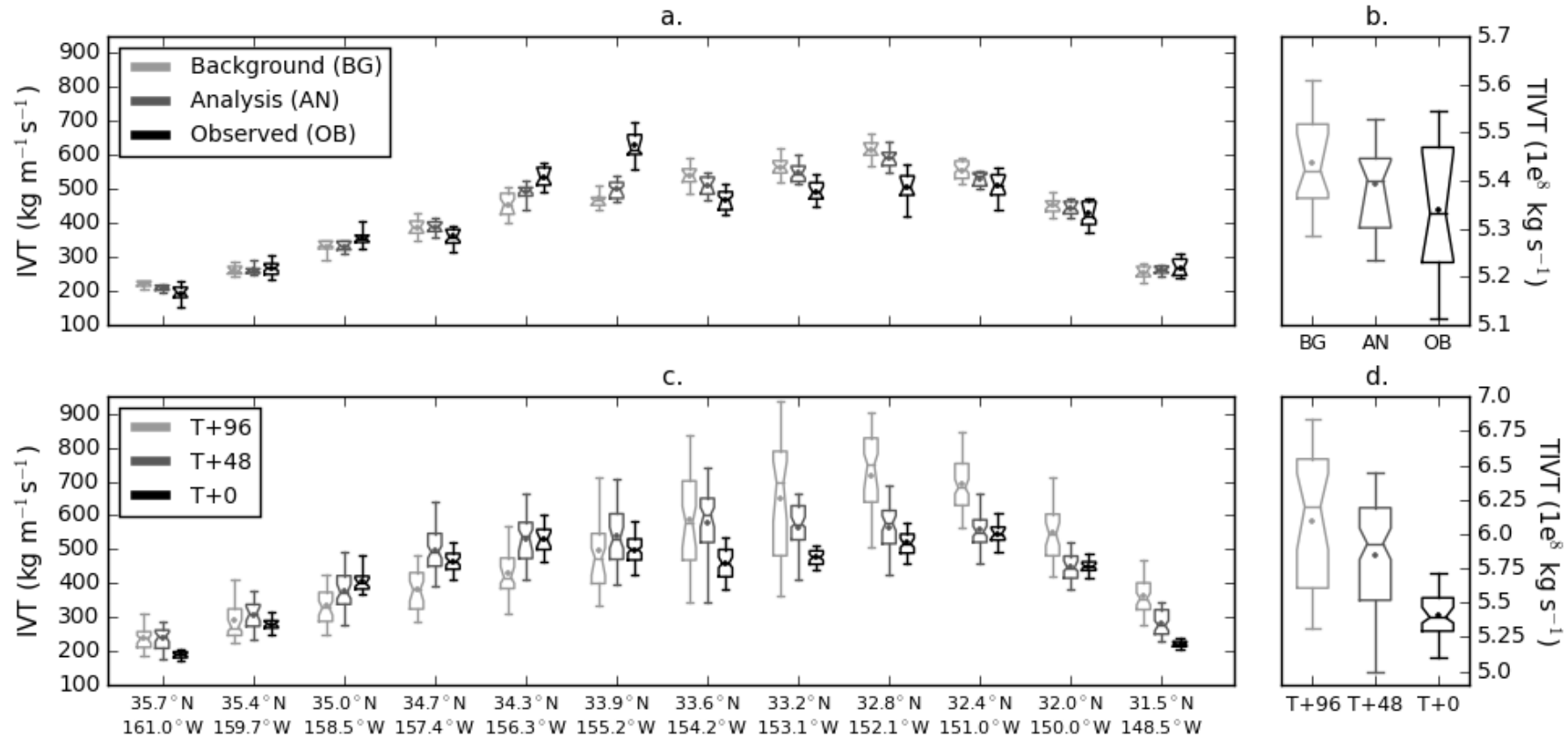
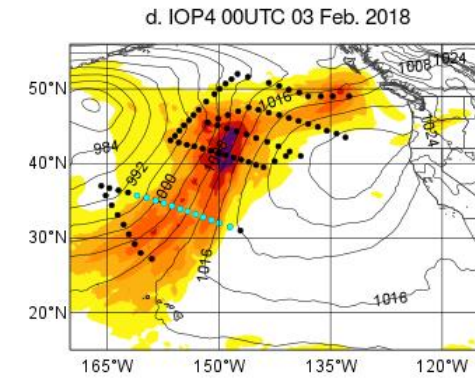
ECMWF data coverage (used observations) - RADIOSONDE

27/01/2018 00

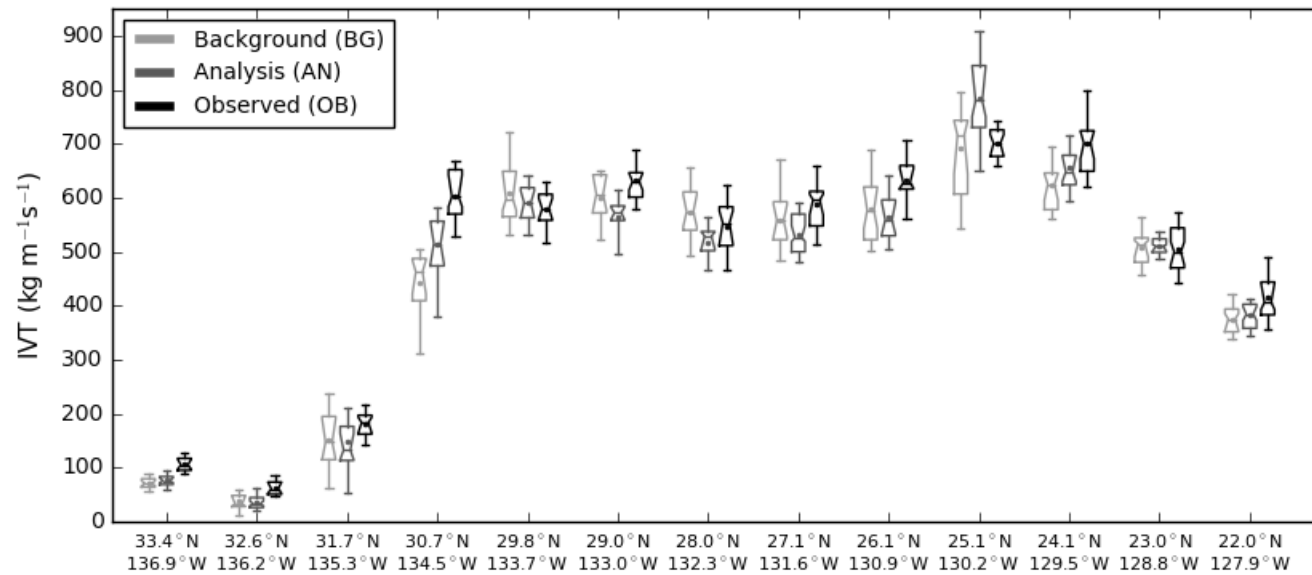
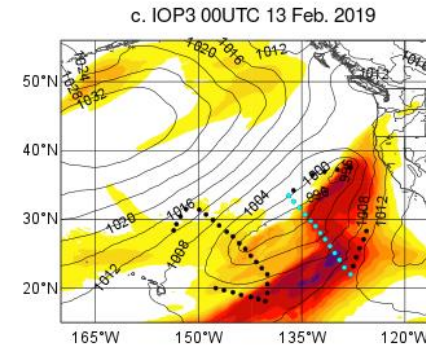
Total number of obs = 864



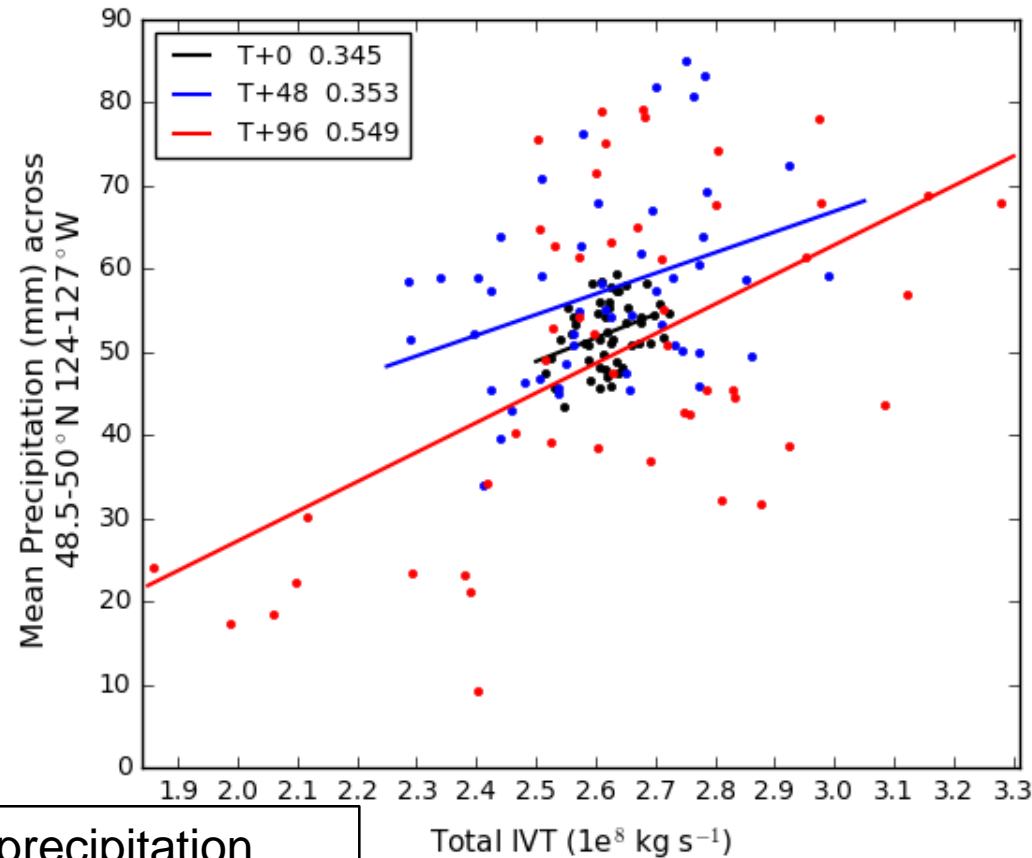
An AR transect in 2018



An AR transect in 2019

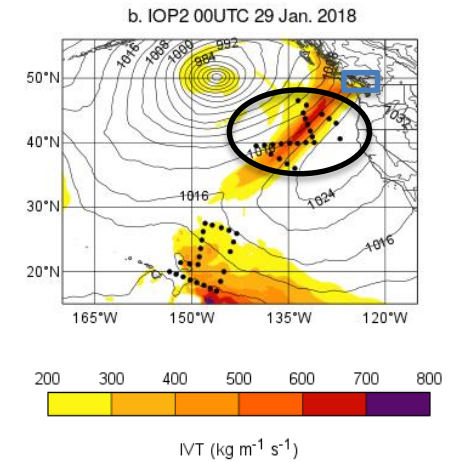


Precipitation uncertainty



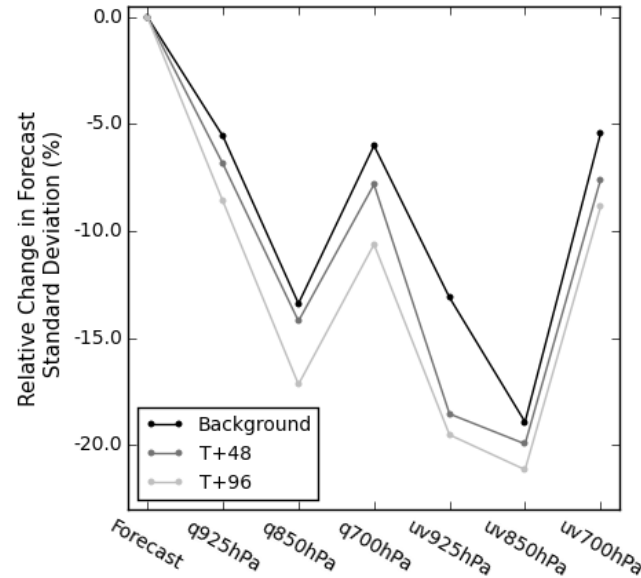
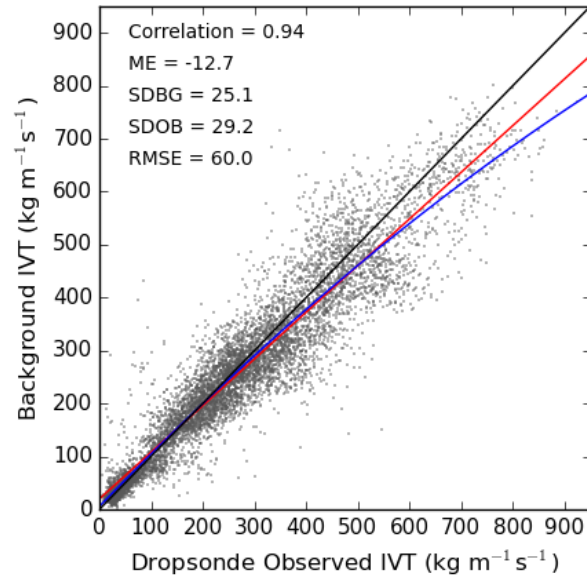
Mean model precipitation
across southwestern
Vancouver Island (48.5°N-
50°N 124°W-127°W)

Total IVT flux valid at 00UTC 29th January
2018 averaged over the four transects

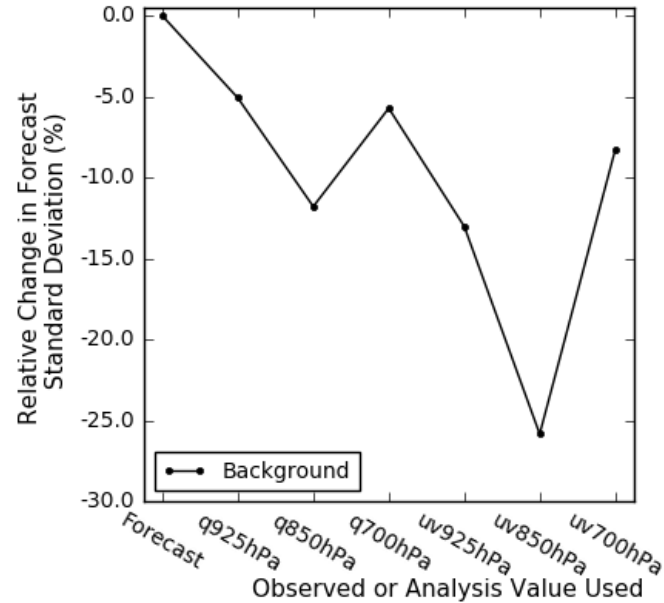
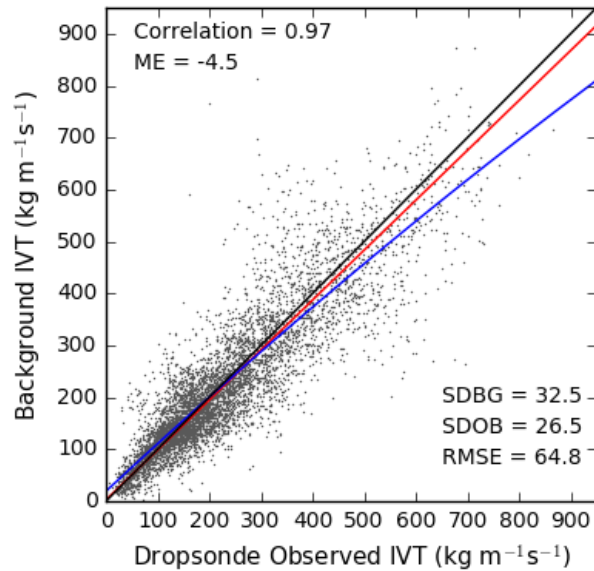


Water vapour flux (IVT) uncertainty

2018



2019

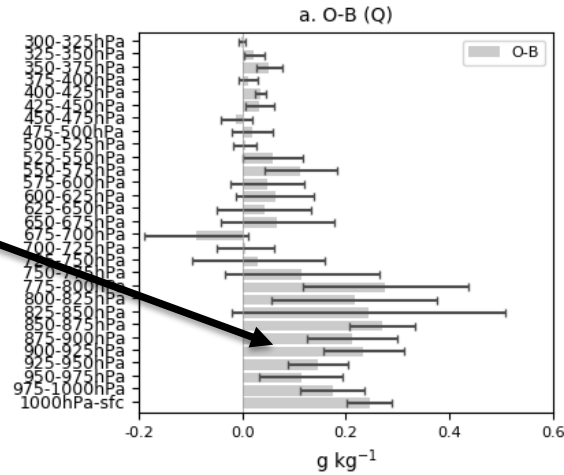


Analysis – part 2

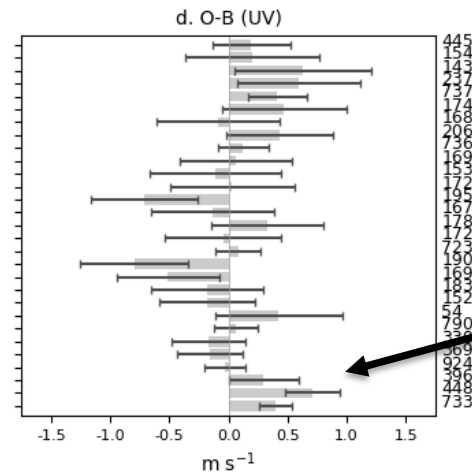
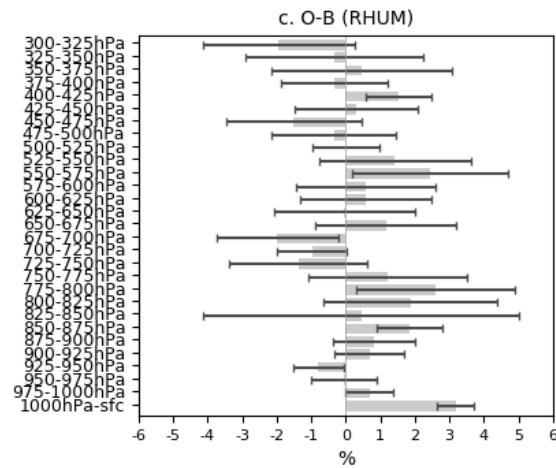
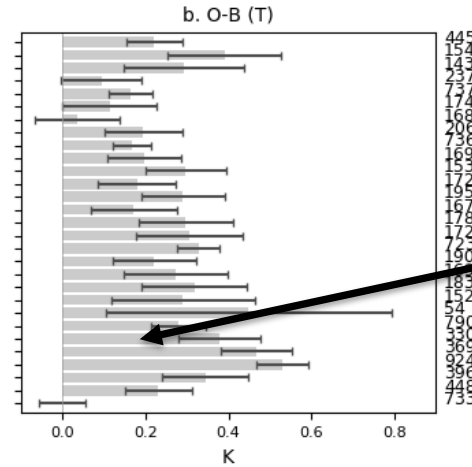
- ECMWF ensemble of data assimilations (EDA).
- Use all available assimilated pressure levels.
- Calculate average observation – background (O – B) statistics in 25hPa layers.
- Assess specific and relative humidity, temperature, and winds.
- Investigate pressure level water vapour fluxes.

Observation – Background (O – B)

Dry bias - observations have higher specific humidity.

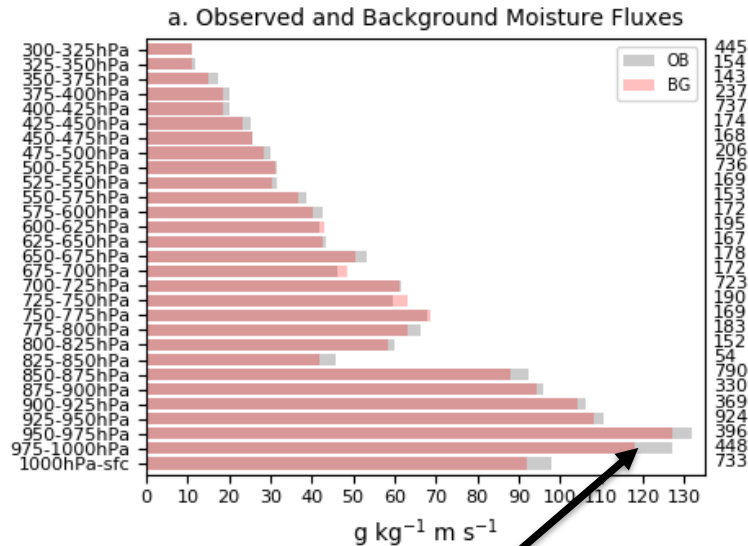


Cold bias - observations are warmer than the model.

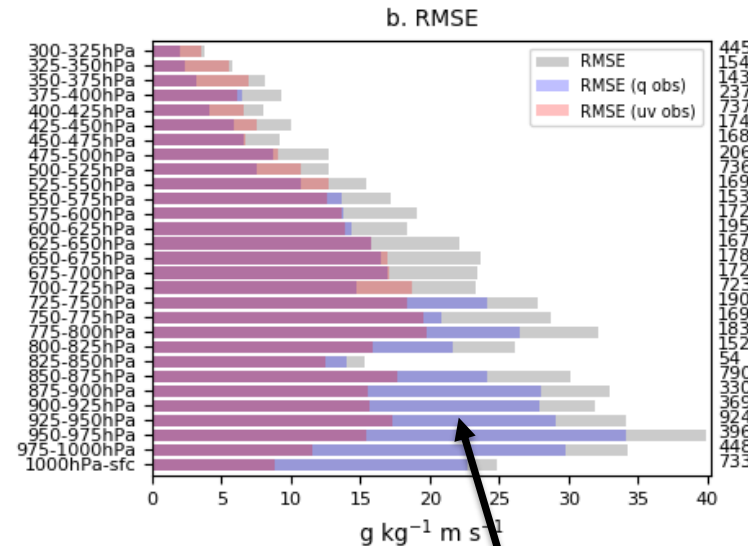


Winds are stronger in the observations in the planetary boundary layer (PBL).

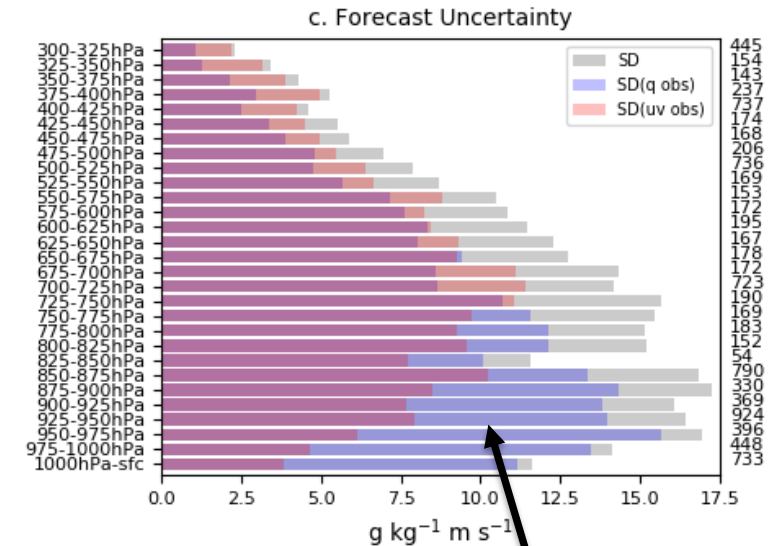
Water vapour fluxes: observed values, RMSE, and uncertainty



Observed water vapour fluxes are higher.



Accurate winds reduce RMSE more than specific humidity at low levels; opposite at high levels.



Accurate winds reduce uncertainty more than specific humidity at low levels; opposite at high levels.

Conclusions – Part 1

- 11 IOPs during AR Recon 2018 and 2019.
- AR structure and IVT magnitude generally well captured.
- Using few pressure levels (925, 850, 700 hPa), uncertainties in winds at 850 hPa cause the most IVT uncertainty.
- Specific humidity is also subject to relatively large uncertainties.

Conclusions – Part 2

- Using all pressure levels, a dry and cold bias is found in the model.
- Observed winds are stronger than the model winds in the planetary boundary layer.
- Water vapour fluxes are underestimated throughout the atmosphere.
- Winds cause most error and uncertainty at lower levels; specific humidity causes more error and uncertainty at higher levels.
- New observations (e.g. from satellite) may improve initialisation of these variables.