

# Statistical Analysis for High Resolution Data Assimilation: Falklands Campaign 2000/2001

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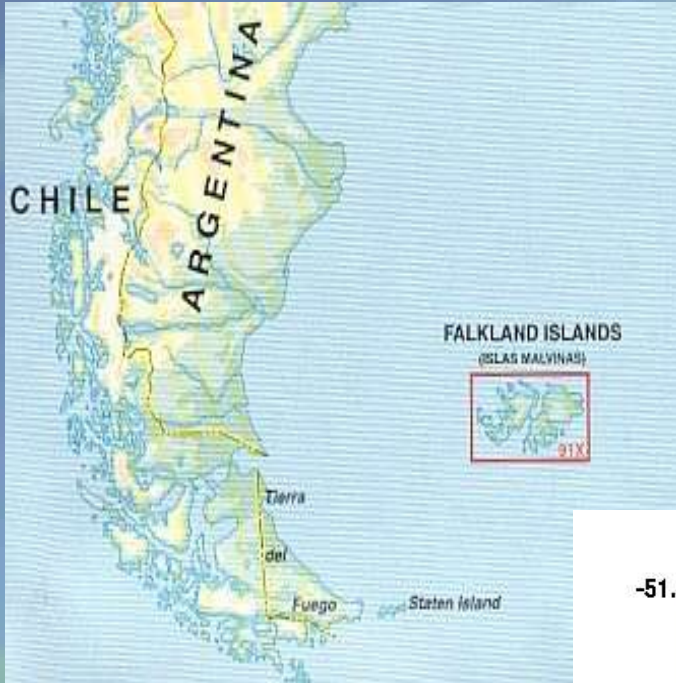
- Introduction
- 3d VOM
- Key findings from statistical analysis
- Implementation into 3d VOM
- Summary

# Mount Pleasant Airport – East Falkland Island

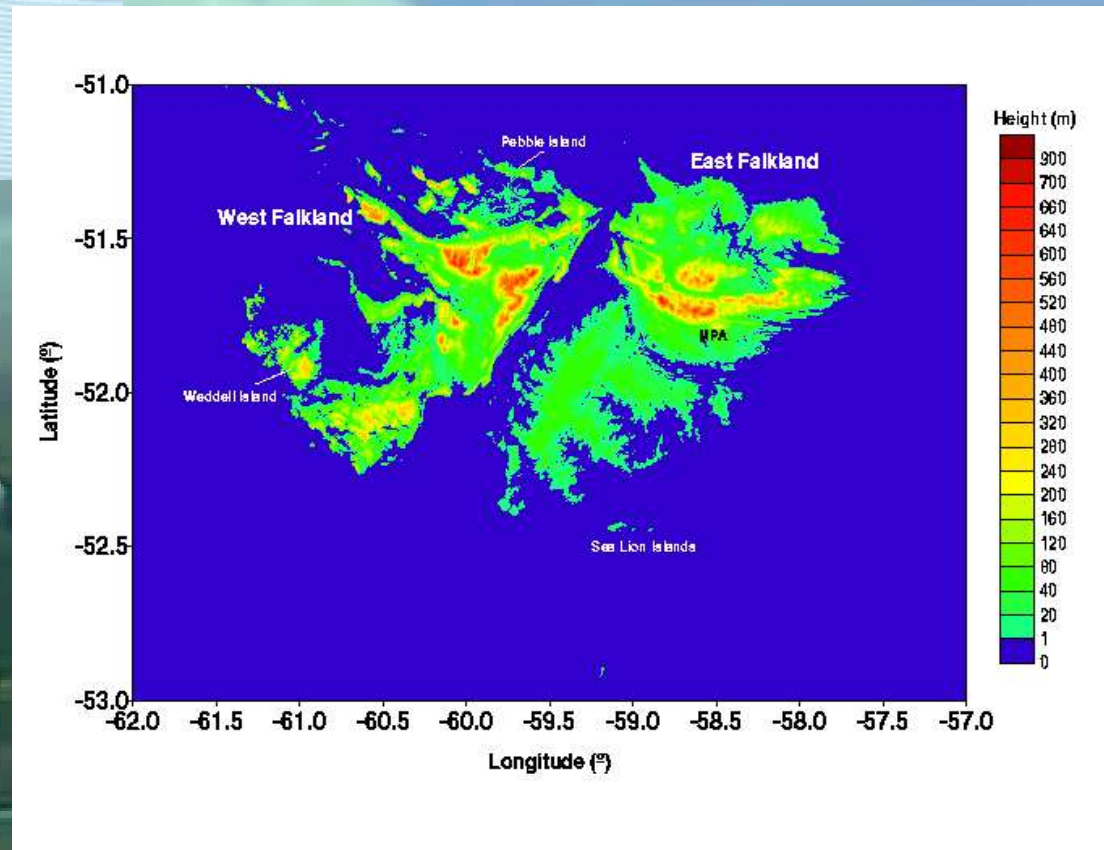


- 3d VOM model developed by Simon Vosper of Met Office/University of Leeds
- Accuracy limited by lack of sophistication in boundary layer processes
- Better understanding of these together with development of data assimilation scheme offers scope for performance enhancement
- No currently operational scheme viable for application to this model

# Falklands Campaign - 2000/2001



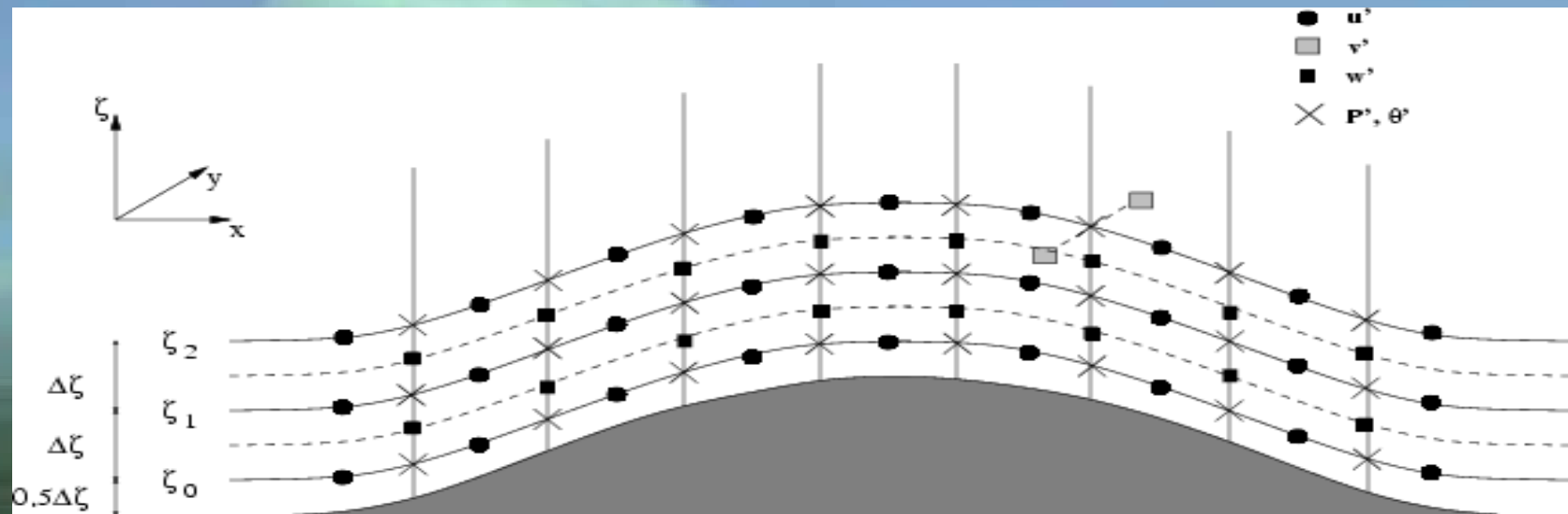
- 20 Automatic weather stations
- Majority clustered around airport
- Remainder in rough north-south transect of island
- Logged data at 30 sec averaged intervals
- Experiment ran for a full year



- Introduce observationally derived nudging constraints to model
- Reasonably straightforward as 3D Vom is a linear model  $\therefore$  no requirement for calculation of adjoint matrix
- Nudging terms in model equation present in each iteration of model run – not simply a case of adjusting initialisation step
- Model will run to alternative steady state

# 3d VOM Schematic

- Terrain following
- Linear model for flow over hills
- Initialised with upwind radiosonde data
- Real orography data set
- 100m resolution in  $x$  and  $y$



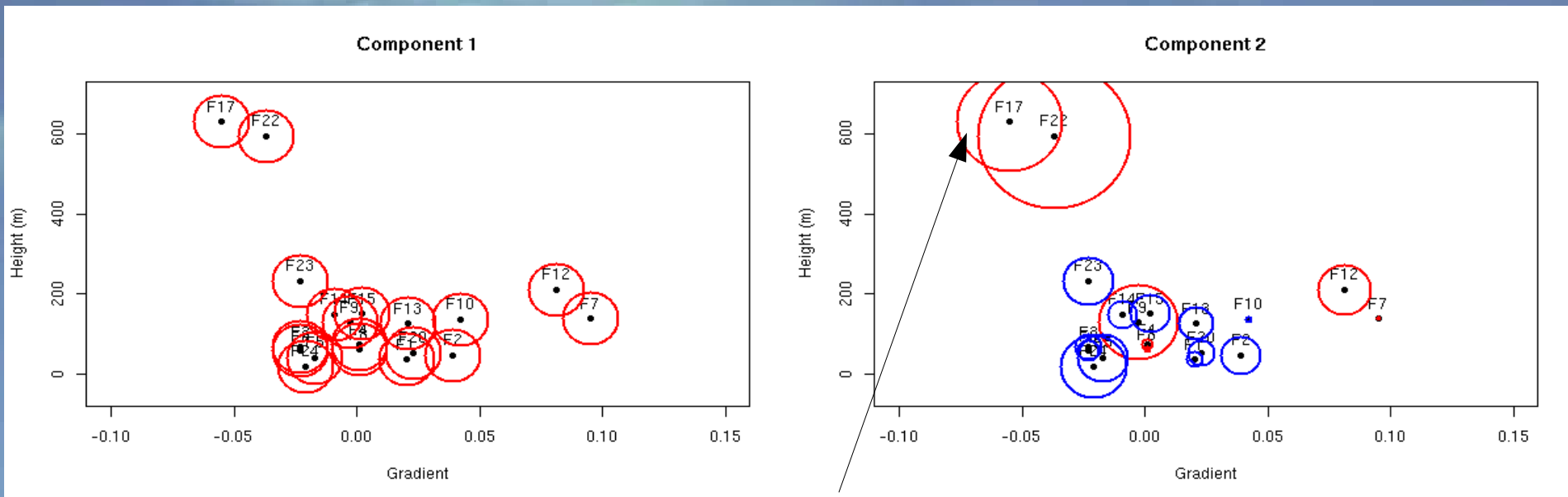
Schematic diagram illustrating the computational grid and the location of the variables on the staggered mesh. Lowest grid level occurs at one half of a grid length above the hill surface.

- To facilitate assimilation step, must understand spatio-temporal behaviour of near surface data
- High-density temporal data yields strong station-wise autocorrelation (  $AR(p)$  processes)
- Spatially sparse data, experimental design and turbulent system means no obvious spatial correlation between observation sites
- Standard interpolation/Kriging techniques not feasible
- PCA (aka EOF) technique applied to observational data to help determine underlying spatial/temporal patterns

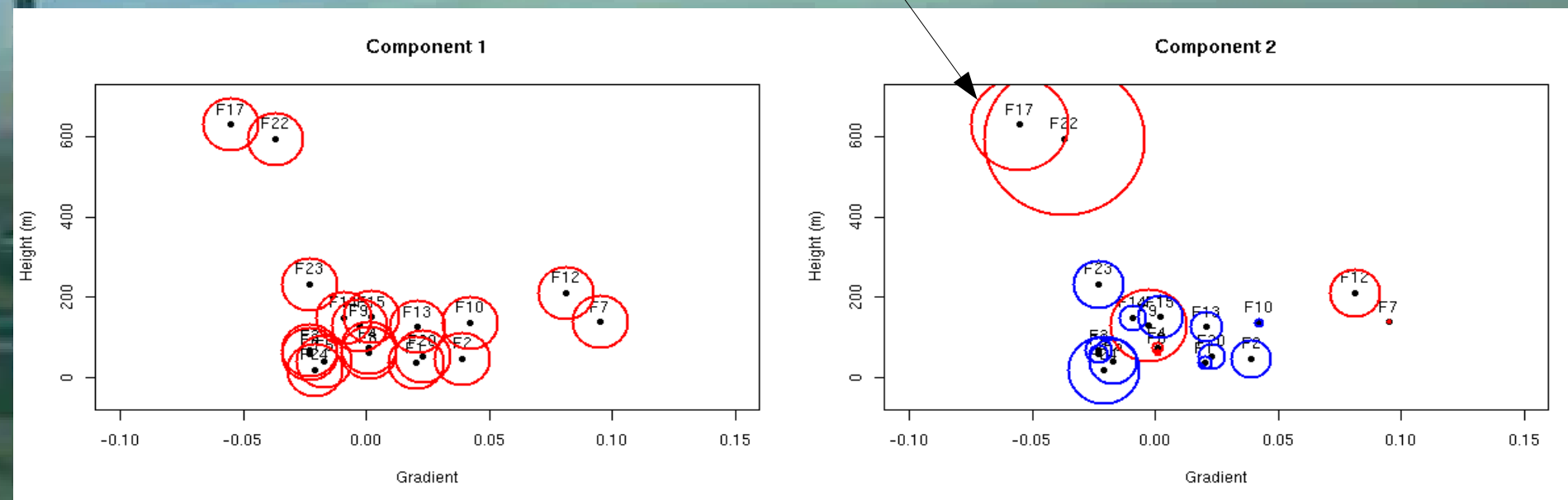
# Principal Component Analysis

- Objective method for determining underlying trends in data
- May be used as data reduction technique whilst retaining most of variability in data set
- Examines variance associated with principal orthogonal components of a time series in both spatial and temporal formats
- Components representing small proportion of cumulative variance can still provide great insight
- Provides an alternative to the fast Fourier Transform

# Raw Data – 10 minute averaged

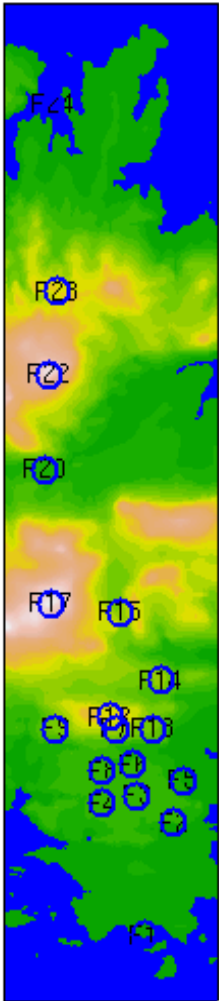


# Hydrostatically Corrected Data – 10 minute averaged

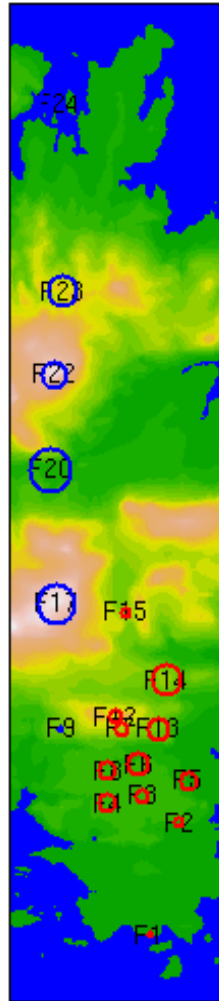


# Case Study: May 17 – no rotor activity

may17 EOF 1



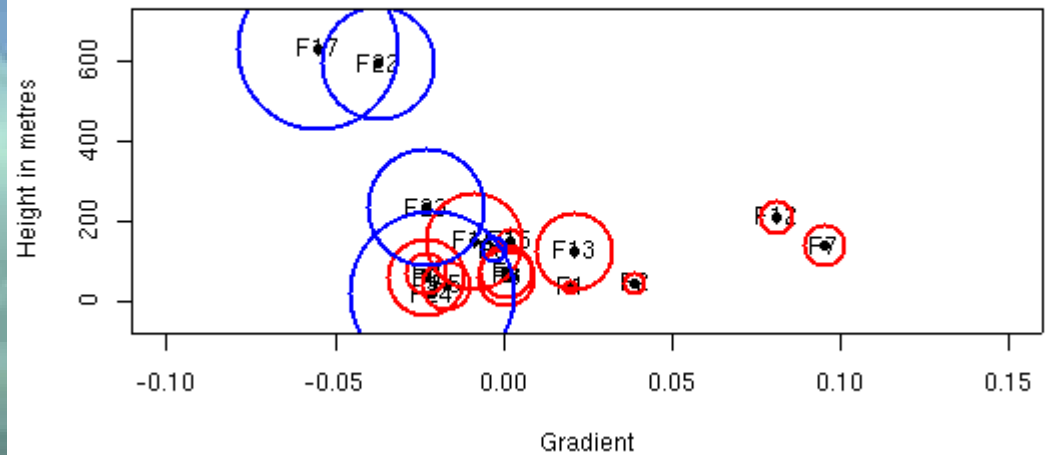
may17 EOF 2



10 min vs 3min almost identical

EOF 1: 98.5% of variance  
Constant throughout field – indicative of steady synoptic scale flow

May17 - 3 min data: EOF 2

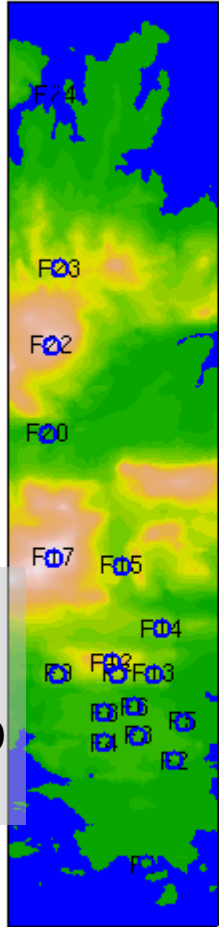


EOF 2: 1.0% of variance  
On first inspection appears driven by east/west split. Mapping onto alternative axes reveals driven by gradient. Represents the drag acting on the system.

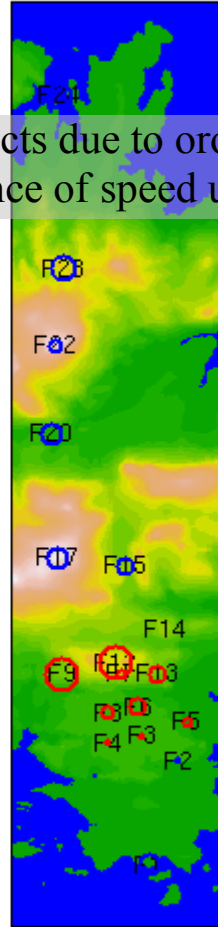


# Case Study: Feb 09 – rotors, downslope windstorms

feb09 - 10min EOF 1



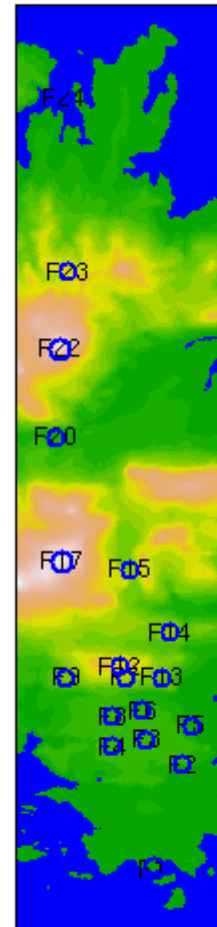
feb09 - 10min EOF 2



Drag effects due to orography superseded by evidence of speed up around airfield

EOF1: 99.5% EOF2: 0.2%

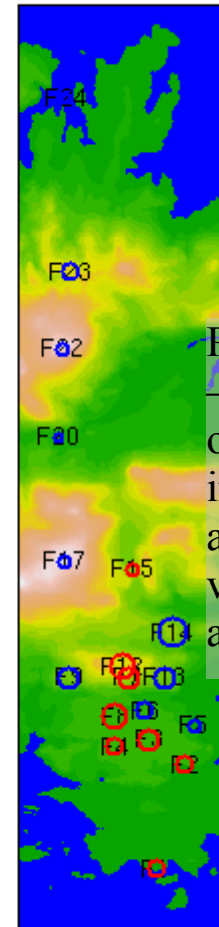
feb09 - 3min EOF 1



Synoptic scale still dominates EOF1 again at 3 min

EOF1: 92.9% EOF2: 4.5%

feb09 - 3min EOF 2



Paired loadings equal with opposite signs indicate wave activity in vicinity of airfield

- Similar analysis performed on 3d VOM model output (*see poster D13*)
- Cross comparison key in providing information regarding position of wave structures – correcting phase shift would dramatically improve model performance
- Assimilation will occur via perturbation of lower boundary conditions
- Model mesh points in close proximity to observation sites will be forced toward observations
- Perturbation relaxed to zero away from region of influence

# Summary

- PCA saves pre-processing of data (negates need for hydrostatic correction)
- Used as diagnostic tool – threshold values for EOF1 (~95%) used to identify turbulent events
- Even when representing a small proportion of cumulative variance, can still be extremely significant
- EOF2 in non-turbulent conditions represents drag on system – opportunity for spatial modelling
- EOF2 gives strong indication of speed up and wave propagation - “fingerprint” for rotor events
- PCA of different time averaged data sets gives insight into temporal variation of events
- Combine observational PCA with model output PCA to facilitate assimilation step

Thank you!

