Sources of predictability and error in ECMWF long range forecasts

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- Overview of System 4
- Some recent research results



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Seasonal prediction at ECMWF

- Started in the 1990's
- Strategy: fully coupled global GCMs

Real-time forecasts since early 1997

○ Forecasts issued publicly from December 1997

Now using "System 4"

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○ Lifetime of systems has been about 5 years each

S1	S2	S 3	S 4	
Dec	Mar	Mar	Nov	
1997	2002	2007	2011	



System 4 seasonal forecast model

IFS (atmosphere)

- \bigcirc T_L255L91 Cy36r4, 0.7 deg grid for physics (operational in Dec 2010)
- Full stratosphere, enhanced stratospheric physics
- Singular vectors from EPS system to perturb atmosphere initial conditions
- Ocean currents coupled to atmosphere boundary layer calculations

NEMO (ocean)

Global ocean model, 1x1 resolution, 0.3 meridional near equator
NEMOVAR (3D-Var) analyses, newly developed.

Coupling

- Fully coupled, no flux adjustments
- Sea-ice based on sampling previous five years

Reduced mean state errors

T850

U50





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Tropospheric scores

Spatially averaged grid-point temporal ACC

ACC S3 and S4 (m2-4; 30y)



ACC S3 and S4 (m5-7; 30y)









More recent ENSO forecasts are better



System 4 configuration

Real time forecasts:

○ 51 member ensemble forecast to 7 months

○ SST and atmos. perturbations added to each member

15 member ensemble forecast to 13 months

- Designed to give an 'outlook' for ENSO
- Only once per quarter (Feb, May, Aug and Nov starts)

Back integrations from 1981-2010 (30 years)

○ 15 member ensemble every month

○ 15 members extended to 13 months once per quarter



How many back integrations?

Back integrations dominate total cost of system

- System 4: 5400 back integrations (must be in first year)
 - 612 real-time integrations (per year)

Back integrations define model climate

- Need both climate mean and the pdf, latter needs large sample
- May prefer to use a "recent" period (30 years? Or less??)
- System 2 had a 75 member "climate", S3 had 275, S4 has 450.
- Sampling is basically OK

Back integrations provide information on skill

- A forecast cannot be used unless we know (or assume) its level of skill
- Observations have only 1 member, so large ensembles are less helpful than large numbers of cases.
- Care needed e.g. to estimate skill of 51 member ensemble based on past performance of 15 member ensemble

O For regions of high signal/noise, System 4 gives adequate skill estimates
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For regions of low signal/noise (eg <= 0.5), need hundreds of years



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Problematic ozone analyses



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Stratospheric trends



Stratospheric temperature trend problem. This is due to an erroneous trend in initial conditions of stratospheric water vapour.





Land surface



Snow depth limits, 1st April



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Sea ice











40'W 120'W 100'W 80'W 60'W 40'W 20'W 0'E 20'E 40'E 80'E 80'E 100'E 120'E 140'E 160'E

CECMWF

sor a

ace a

70* S

© ECMWF

50°S

60°S

7019

Model errors are still serious ...

Models have errors other than mean bias

- Eg weak wind and SST variability in System 2
- Past models underestimated MJO activity (S4 better)
- Suspected too-weak teleconnections to mid-latitudes

Mean state errors interact with model variability

- Nino 4 region is very sensitive (cold tongue/warm pool boundary)
- Atlantic variability suppressed if mean state is badly wrong

• Forecast errors are often larger than they should be

- With respect to internal variability estimates and (occasionally) other prediction systems
- Reliability of probabilistic forecasts is often not particularly high (S4 better)



Recent Research



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S4 extended hindcast set

15 members

51 members



(Figures from Susanna Corti)



S4 extended hindcast set

15 members

51 members



(Figures from Susanna Corti)



S4 ACC DJF Z500

Anomaly Correlation Coefficient for ECMWF S4 with 51 ensemble members 500 hPa geopotential height

Hindcast period 1981-2010 with start in November average over months 2 to 4 Black dots for values significantly different from zero with 95% confidence (1000 samples)

S4 ACC perfect model limit

Perfect-model Anomaly Correlation Coefficient for ECMWF S4 with 51 ensemble members 500 hPa geopotential height

Hindcast period 1981-2010 with start in November average over months 2 to 4 Black dots where perfect model assumption is violated with 95% confidence (1000 samples)







0.0 1.911336+45

Local p-value for perfect model

p-value for observed ACC, assuming perfect model for ECMWF S4 with 51 ensemble members 500 hPa geopotential height

Hindcast period 1981-2010 with start in November average over months 2 to 4

p-value for observed ACC, assuming perfect model for ECMWF S4 with 51 ensemble members Mean sea level pressure

Hindcast period 1981-2010 with start in November average over months 2 to 4



Indistinguishable from perfect Worse than perfect Better than perfect



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Fig. 1. The Arctic Oscillation Index for DJF, as analysed from ERAI (blue) and as predicted by the S4 ensemble mean from the 1st November (red). The S4 ensemble mean is **scaled by a factor of 6** to be of comparable amplitude to the observed index.





A big reduction in vertical diffusion, and a further tuning of non-orographic GWD, has given a big additional improvement in the QBO compared to S4.



Period and downward penetration match observations Semi-annual oscillation still poorly represented



QBO forecasts



Magics 2.18.0 (64 bit) - nautilus - net - Mon Oct 22 12:26:23 2012

CECMWF



S3

S4

New

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NH winter forecasts: vertical diffusion





Z500 Anom. correlation fuhg(101)-ERA-Int 1981-2010DJF Global z-mean acc: 0.666 NH:0.371 TR:0.829 SH:0.403



Z500 Anom. correlation fulf(101)-ERA-Int 1981-2010DJF Global z-mean acc: 0.664 NH:0.319 TR:0.832 SH:0.424



Fisher z transform diff fuhg(101)-fulf(101) 1981-2010DJF sigma: 0.272 mean: 0.004



0.371

0.319

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NH winter forecasts

MSLP Anom. correlation fuhg(101)-ERA-Int 1981-2010DJF Global z-mean acc: 0.575 NH:0.339 TR:0.729 SH:0.381



MSLP Anom. correlation fulf(101)-ERA-Int 1981-2010DJF Global z-mean acc: 0.563 NH:0.279 TR:0.72 SH:0.413



Fisher z transform diff fuhg(101)-fulf(101) 1981-2010DJF sigma: 0.272 mean: 0.0175





MSLP Ens. mean S/N ratio fuhg(101)-ERA-Int 1981-2010DJF

Even with 101 members, ensemble mean signal not always well defined



135°E

45°E 90°E 45°₩

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60°I



Conclusions

Models are improving

- Gradual but continuous improvement in scores
- Reliability can be high in many situations

• Forecast systems still have deficiencies

- Need calibration, and often cannot be trusted at face value
- Some issues may affect real-time forecasts more than re-forecasts

• Further improvements lie ahead

- Research results suggesting that previous estimates of predictability limits might be *wrong*.
- Hard work needed to improve models and capture new sources of predictability.



