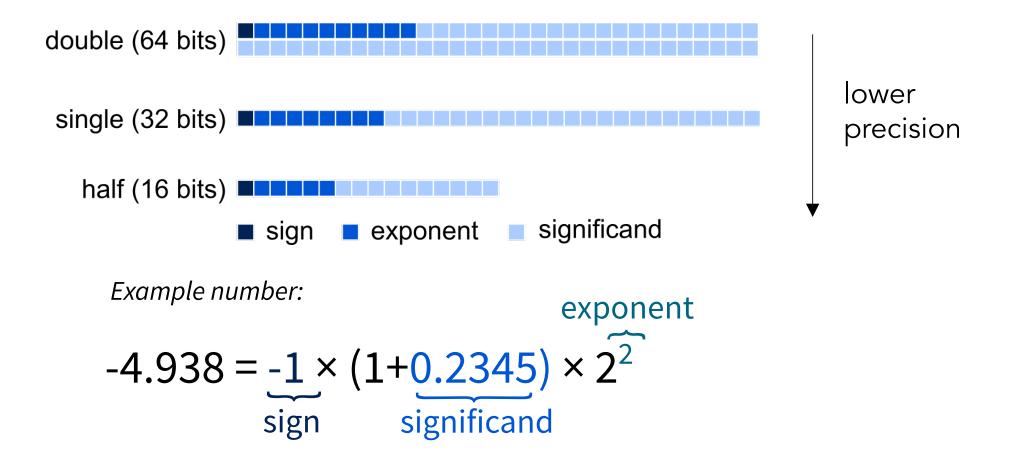


# Reducing Precision in Ensemble Data Assimilation

Sam Hatfield, Peter Düben, Matthew Chantry, Tim Palmer (also Aneesh Subramanian, Keiichi Kondo, Takemasa Miyoshi) samuel.hatfield@physics.ox.ac.uk

# Floating point arithmetic

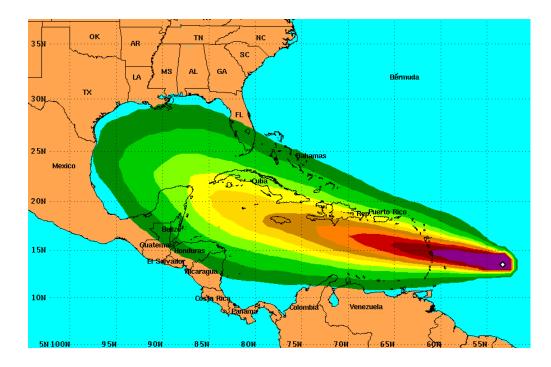


## Why reduce numerical precision?

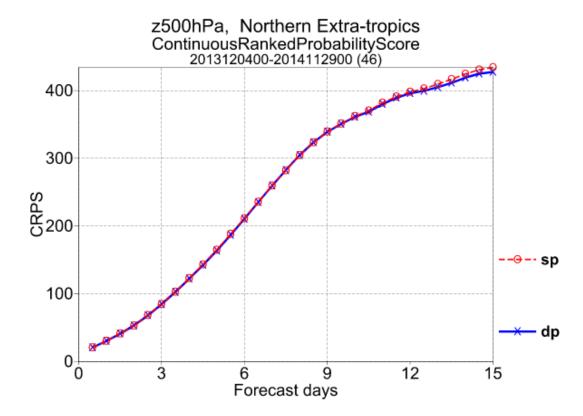
### 1) Computing trends



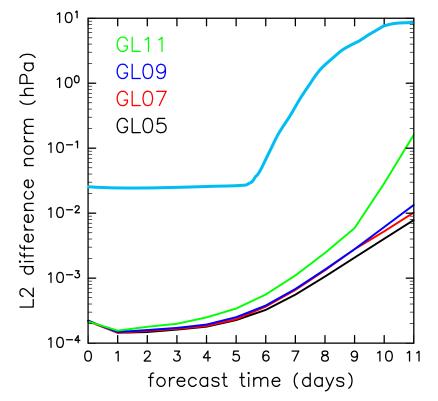
#### 2) Model uncertainties



### Single precision in NWP



ECMWF's Integrated Forecasting System (Váňa et al., 2016, MWR) Jablonowski-Williamson benchmark (QJRMS 2006)



NICAM (non-hydrostatic, icosahedral) (Nakano et al., 2018, MWR)

## Below single precision

QUADRO



#### Journal of Advances in Modeling Earth Systems

#### RESEARCH ARTICLE

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#### Key Points:

 Huge saving in computing cost via reduced numerical precision in earthsystem modeling
 Long and short-term simulations have similar level of minimal numerical precision
 Numerical precision can be reduced with time in a weather forecast simulations

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### On the use of programmable hardware and reduced numerical precision in earth-system modeling

2015MS000494.pdf (page 1 of 16)

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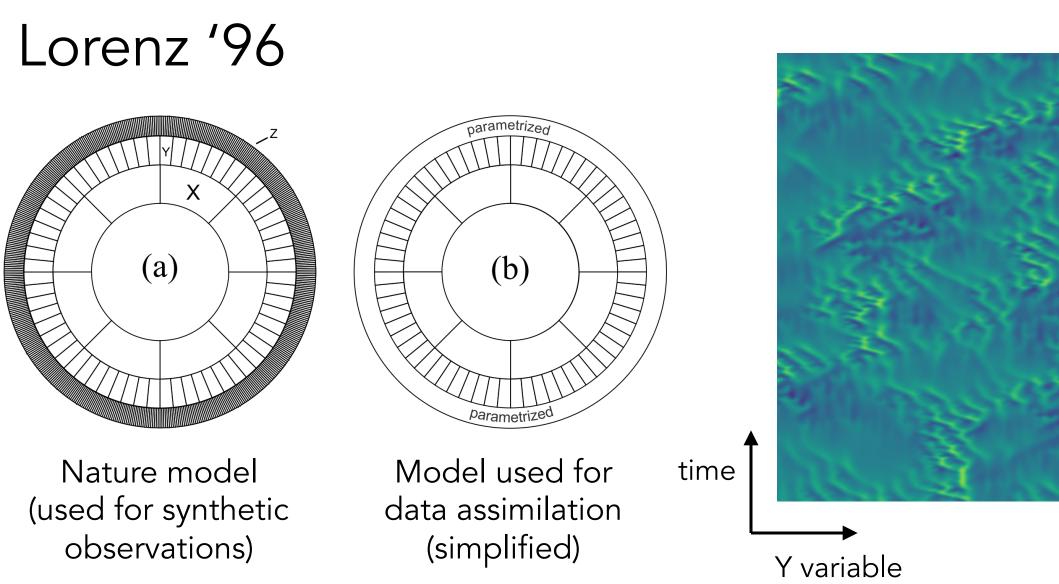
**Abstract** Programmable hardware, in particular Field Programmable Gate Arrays (FPGAs), promises a significant increase in computational performance for simulations in geophysical fluid dynamics compared with CPUs of similar power consumption. FPGAs allow adjusting the representation of floating-point numbers to specific application needs. We analyze the performance-precision trade-off on FPGA hardware for the two-scale Lorenz '95 model. We scale the size of this toy model to that of a high-performance computing application in order to make meaningful performance tests. We identify the minimal level of precision at which changes in model results are not significant compared with a maximal precision version of the model and find that this level is very similar for cases where the model is integrated for very short or long intervals. It is therefore a useful approach to investigate model errors due to rounding errors for very short simulations (e.g., 50 time steps) to obtain a range for the level of precision that can be used in expensive long-term simulations. We also show that an approach to reduce precision with increasing forecast time, when model errors are already accumulated, is very promising. We show that a speed-up of 1.9 times is possible in comparison to FPGA simulations in single precision if precision is reduced with no strong change in

# Reducing precision in data assimilation

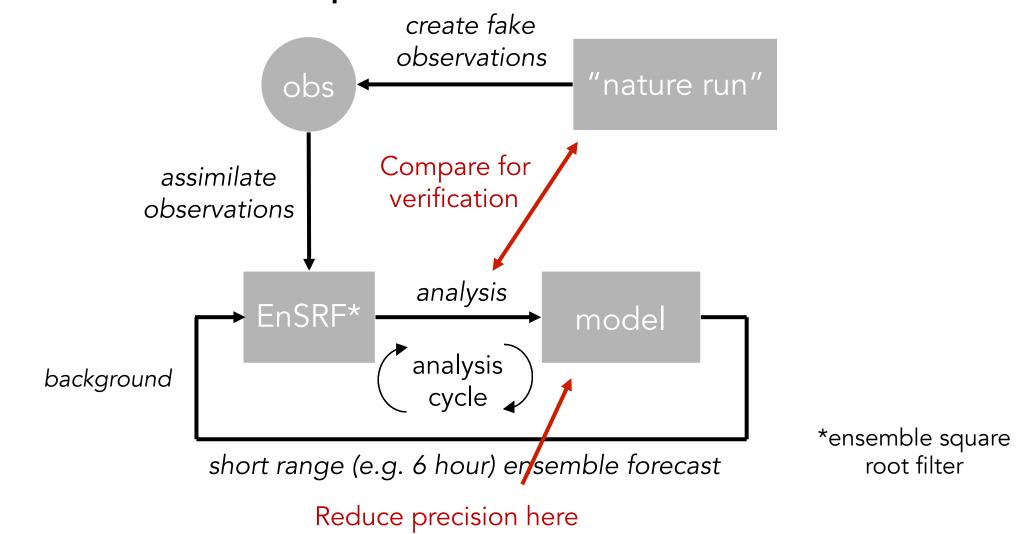
### • Research questions:

- Can we adjust precision to a level justified by "system uncertainty"? (e.g. model error, observation error)
- Can we improve the quality of analyses if we reinvest computational savings from reducing precision? (e.g. boost the ensemble size)
- Talk outline:
  - 1. Lorenz '96/ensemble square root filter
  - 2. SPEEDY/local ensemble transform Kalman filter

### Section 1. Lorenz '96

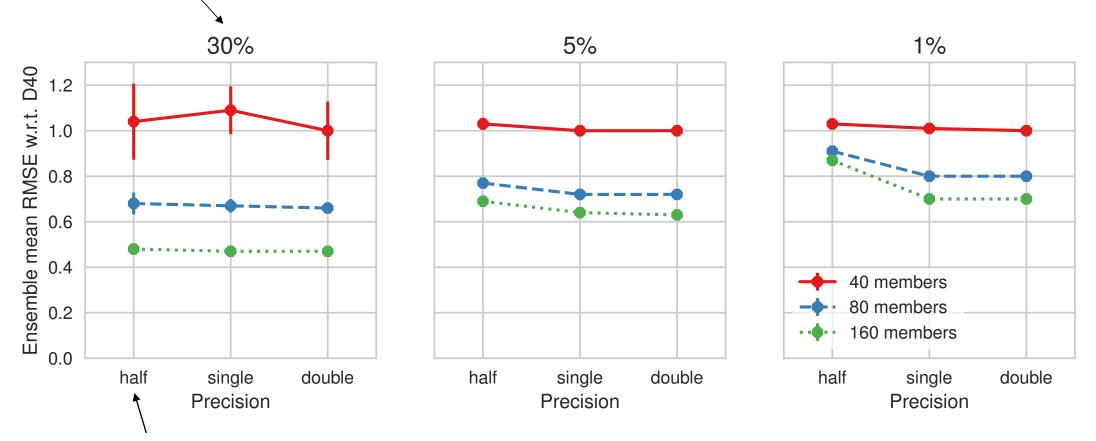


### Assimilation setup



### Reduced precision analyses

Observation error (% of natural variability)



half-precision model used a software emulator

### Precision/ensemble size trade-off

- Trade bits for ensemble members
- Computational speedup from reducing precision is difficult to predict
- But even a modest speed-up can be beneficial

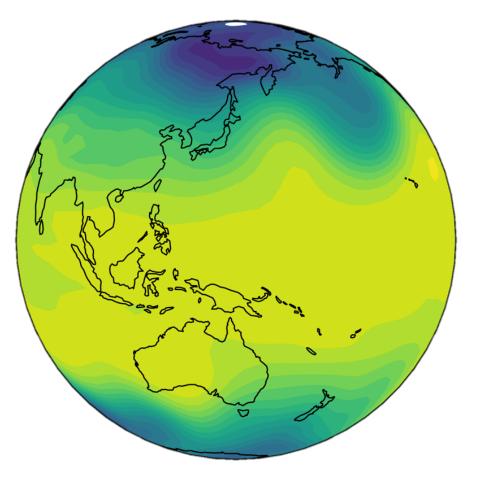
Analysis ensemble mean RMSE w.r.t. D40 1.0 single precision half precision 0.9 0.8 0.7 0.6 0.5 10% 20% 30% 40% 50% 0% 60% 70% 80% Assumed speed up from reducing precision **100**<sup>120</sup> 40 60 160 80 50 Equivalent cost ensemble size

~26% error reduction w.r.t. double precision

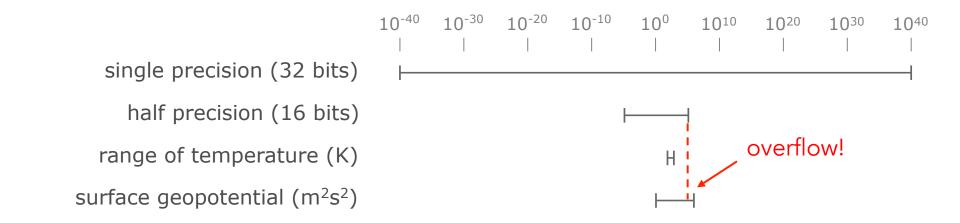
### Section 2. SPEEDY

### SPEEDY

- Spectral dynamical core
- T30 resolution roughly 400km at equator
- Several parametrized processes: convection, radiation, land/sea fluxes
- Reduce precision in forecast model
  → measure change in analysis error

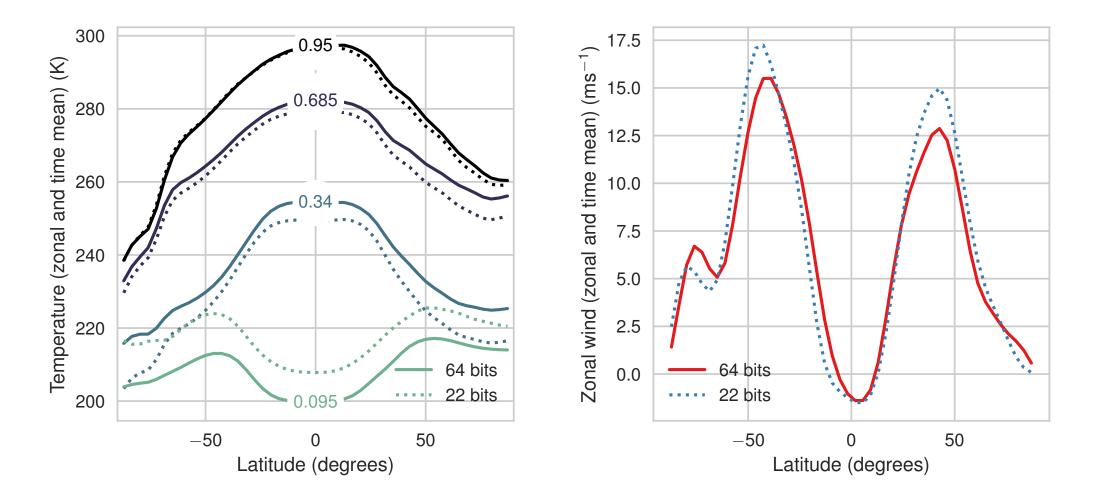


# Using half precision arithmetic in SPEEDY



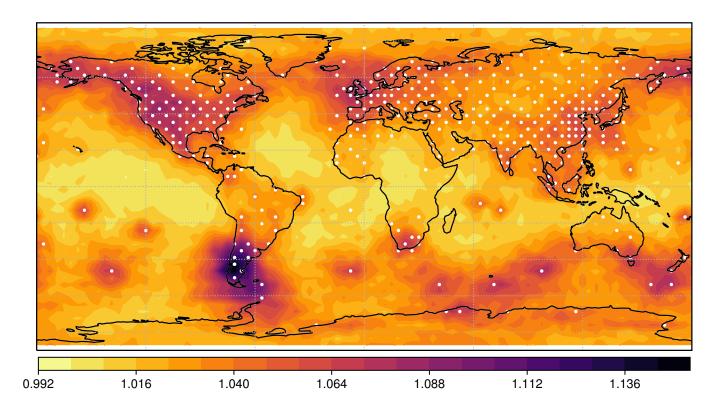
- Half-precision floats have a limited range ( $10^{-5} \sim 10^{6}$ )
- For now, only reduce significand width (52 bits  $\rightarrow$  10 bits)
- Compare 22 bits (1+11+10) with 64 bits (1+11+52) sign exponent significand

### Reduced precision SPEEDY biases



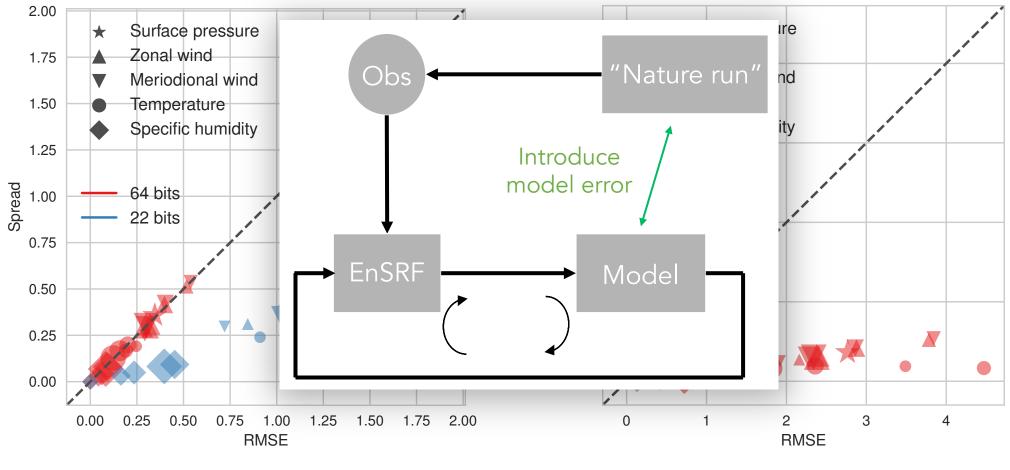
### Assimilation setup

- Assimilation algorithm: local ensemble transform Kalman filter
- Synthetic observations
- 20 members
- Gaspari-Cohn covariance localisation and RTPP inflation



Contours: RTPP inflation factor Dots: observation locations

### Perfect model experiments

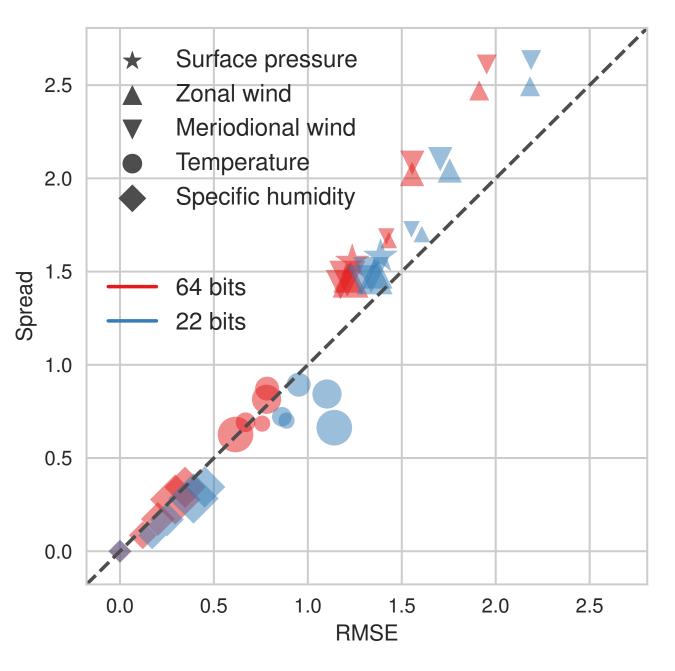


64 bit nature run

22 bit nature run

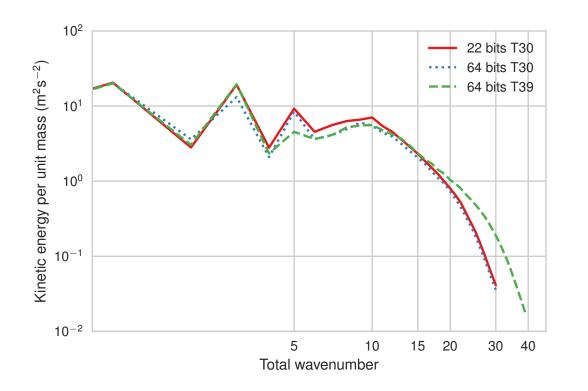
### Introducing model error (1)

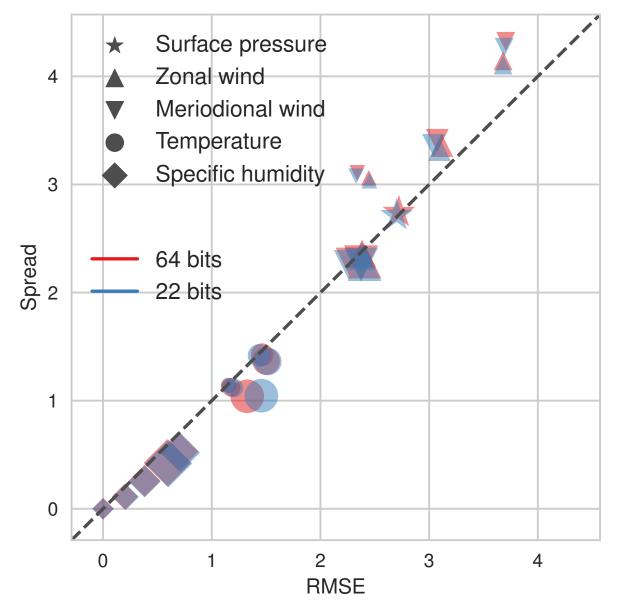
• Try doubling diffusion time scale in assimilation model



### Introducing model error (2)

• What about a higher resolution nature run? (T39 instead of T30)





### Conclusion

- Reducing precision could provide a one-off "boost" of computer resources on the order of a computer upgrade
- The lowest possible precision is constrained by the level of uncertainty (observations, model error etc.)
- The use of half-precision in data assimilation/modeling remains an open question due to the severely limited range

### Time to reconsider IEEE floats?

