

Lawrence Livermore National Laboratory

**Computational Aspects of the
UQ Project at LLNL**

February 14, 2011



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This work performed under the auspices of the U.S. Department of Energy by
Lawrence Livermore National Laboratory under Contract DE-AC52-07NA27344

LLNL-PRES-469892

LLNL UQ Strategic Initiative LDRD.

- Richard Klein (PI), Xabier Garaizar (co-PI)
- Climate Team:
 - Curt Covey (team lead)
 - Donald Lucas (modeling & analysis)
 - John Tannahill (s/w architecture & development)
 - Yuying Zhang (observations & analysis)
- UQ Pipeline Team:
 - David Domyancic
 - Scott Brandon
- A number of other UQ researchers:
 - Gardar Johannesson



CAM/CESM particulars for this work.

- Basic CAM/CESM configuration used:
 - 1.9x2.5° horizontal resolution
 - Finite-Volume dynamical core
 - 26 vertical levels
 - CAM3_5_1 => CAM4 physics
- CAM/CESM namelist code modified to allow for 36 parameter values to be input.
- CESM scripting system modified as needed.
- Extensive Python script developed to insulate user from any CAM/CESM specifics.



CAM/CESM UQ parameters of interest on 2/10.

#	Variable Name	Description	File Name (.F90)	Src*
1	rhminh	Threshold RH for fraction of high stable clouds	cloud_fraction	J
2	rhminl	Threshold RH for fraction of low stable clouds	cloud_fraction	J+S
3	rliqice	Effective radius of liq. cloud droplets over sea ice	pkg_cldoptics	R
4	rliqland	Effective radius of liquid cloud droplets over land	pkg_cldoptics	R
5	rliqocean	Effective radius of liquid cloud droplets over ocean	pkg_cldoptics	R
6	ice_stokes_fac	Scaling factor applied to ice fall velocity	pkg_cld_sediment	S
7	capnc	Cloud particle num. density over cold land/ocean	cldwat	R
8	capnsi	Cloud particle number density over sea ice	cldwat	R
9	capnw	Cloud particle number density over warm land	cldwat	R
10	conke	Evaporation efficiency of stratiform precipitation	cldwat	J
11	icritc	Threshold for autoconversion of cold ice	cldwat	S
12	icritw	Threshold for autoconversion of warm ice	cldwat	S
13	r3lcrit	Critical radius at which autocon. becomes efficient	cldwat	R
14	ricr	Critical Richardson number for boundary layer	hb_diff	K/B
15	c0	<i>Shallow</i> convection precipitation efficiency	hk_conv	J
16	cmftau	Time scale for consumption rate of shallow CAPE	hk_conv	K/B
17	alfa	Initial cloud downdraft mass flux	zm_conv	J
18	c0	<i>Deep</i> convection precipitation efficiency	zm_conv	J
19	dmpdz	Parcel fractional mass entrainment rate	zm_conv	S
20	ke	Environmental air entrainment rate	zm_conv	J
21	tau	Time scale for consumption rate of deep CAPE	zm_conv	J

*Source => **J**:Jackson, C, J Clim 21:6698, '08; **K/B**:Klein, S & Bader, D, Suggestion, '09; **R**:Rasch, P, Suggestion, '09; **S**:Sanderson, B, CCSM Workshop, '09



Added UQ parameters of interest **since 2/10.**

#18 (c0) split,
so only counted
once here.

#	Variable Name	Description	File Name (.F90)	Src*
22	fac	ustar parameter in PBL height diagnosis	hb_diff	Z
23	fak	Constant in surface temperature excess	hb_diff	Z
24	betamn	Minimum overshoot parameter	hk_conv	Z
25	sgh_scal_fac	Land roughness scaling factor	physpkg	T
26	c0_lnd	Deep convec. precipitation efficiency over land	zm_conv	J
	c0_ocn	Deep convec. precipitation efficiency over ocean	zm_conv	J
27	capelmt	Threshold value for CAPE for deep convection	zm_conv	Z
28	cdn_scal_fac	Ocean roughness scaling factor	shr_flux_mod	Z
29	z0m_scal_fac	Mois. & heat resistance to vegetation scaling factor	Biogeophysics1Mod	Z
30	dt_mlt_in	Temperature at which melt begins	ice_shortwave	B
31	r_ice	Sea ice tuning parameter	ice_shortwave	B
32	r_pnd	Ponded ice tuning parameter	ice_shortwave	B
33	r_snw	Snow tuning parameter	ice_shortwave	B
34	rsnw_melt_in	Maximum snow grain radius	ice_shortwave	B
35	ksno	Thermal conductivity of snow	ice_therm_vertical	B
36	mu_rdg	Gives e-folding scale of ridged ice	ice_mechred	B

*Source => B:Bailey,D, etal, Suggestion, '11; J:Jackson, C, J Clim 21:6698, '08; T:Taylor,M, Suggestion '10;
Z:Zhang, M, Suggestion, '10



CAM/CESM UQ runs.

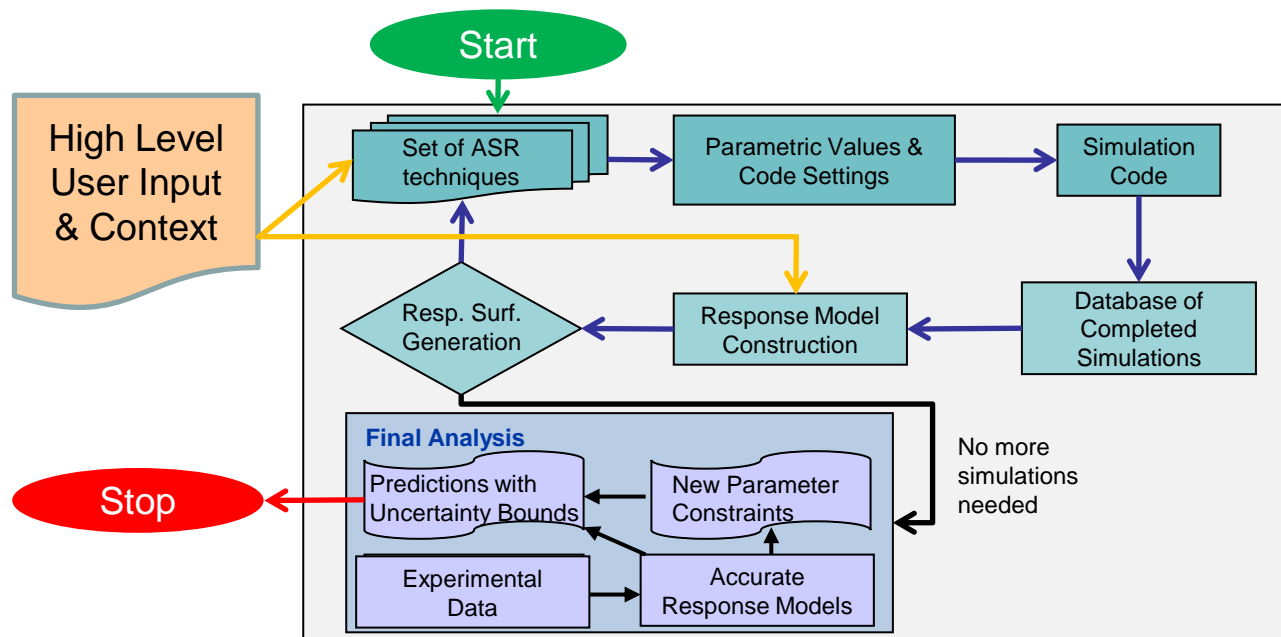
- Run on LLNL's Atlas Linux Cluster.
- Use LLNL's UQ Pipeline tool to automate runs.
 - Python interface scripts developed.
- Runs to date:

CAM/CESM version	ocn mode	# params	run dates	# runs	simyrs/run	total simyrs	tasks/run	thrs/task	cores/run	concur. runs	cores/job
CAM3.6	AMIP	21	1/10-5/10	1,242	12	14,904	192	2	384	11	4,224
CAM4	AMIP	21-28	5/10-1/11	1,695	12	20,340	192	2	384	11	4,224
CESM1	AMIP	29	1/11	59	12	708	192	2	384	11	4,224
CESM1	SOM	36	2/11-	Test	30-60?	Test	192	4	768	4	3,072
Total				2,996		35,952					



LLNL UQ Pipeline tool.

- Stages & executes a set of concurrent ensemble simulations.
- Provides a wide range of sampling strategies, as well as analysis capabilities like “MARS”.
- Self-guiding, self-adapting technologies are being developed that will automatically & efficiently steer the study parameter space to explore.



LLNL CESM/UQ system is fully automated.

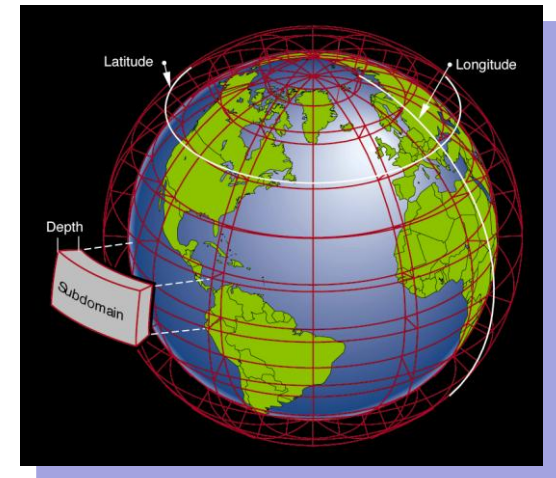
- “Push the button” & in theory you can do a study made up of 100’s of runs without intervening.
- Study is relatively simple to set up; 3 files required:

- `uq_info.py`:
 - Contains pipeline/application interface data dictionaries.
 - Generally modify a couple of items here at the beginning of each study.
- `appl_interface.py`:
 - Provide `prep_ensemble`, `prep_run`, `post_run`, & `post_ensemble` interface functions required by pipeline.
 - Generally no need to modify once established.
- `lcesm_run.py`:
 - Shields user from CESM details.
 - Generally no need to modify.



Migration from CAM standalone to CESM.

- More difficult than anticipated.
- “UQ ensemble runs” does not appear to be a use case that the CESM scripts currently facilitate?
- The primary CESM setup functionality that is needed for our work is just:
 - Configure/Build the code.
 - Set up the namelist files.
- Many thanks to the various NCAR people who helped with this transition.



“Atlas Node 587” issue – Discovered by chance.

- Inadvertently deleted some files from a study & had to redo some runs.
- Old summary diagnostic files should have matched the new ones “bit for bit”, but some did not?
- Ended up taking ~2 months to resolve this issue.
 - A concerted effort by a number of people.



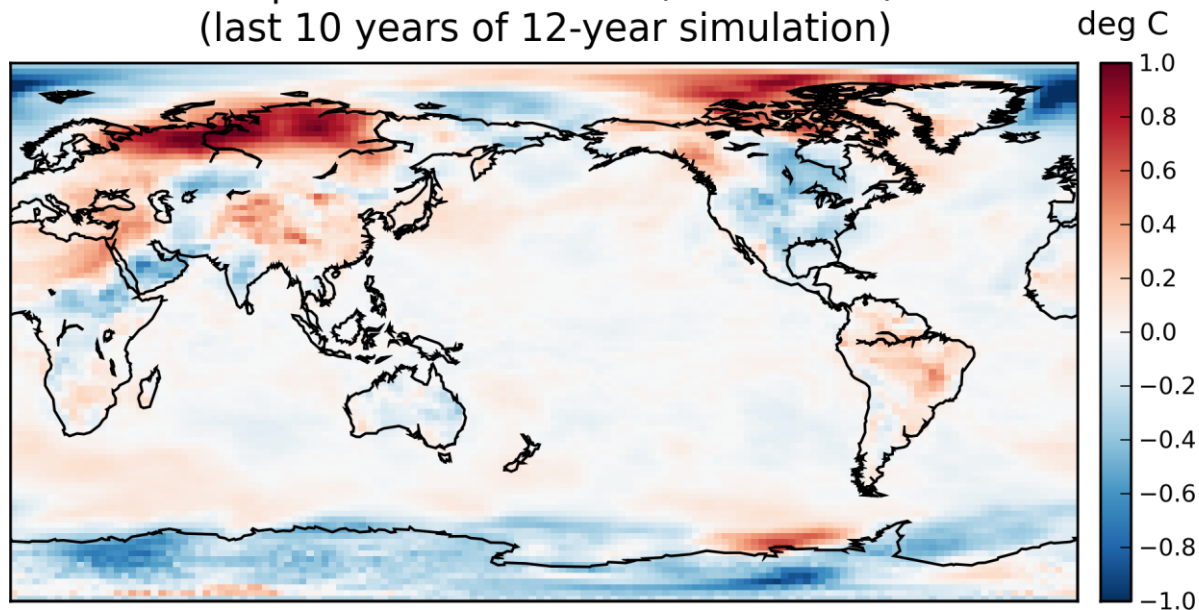
“Atlas Node 587” issue – The symptoms.

- Made many many test runs, which eventually led to the following observations:
 - Atlas runs using Node 587 could lead to “some sporadic bad diagnostic output”.
 - Bad output values were still “plausible”.
 - Bad output values were “significantly different” from what they should have been.
 - Atlas runs not using Node 587 produced no bad output.
 - Running on a different, but very similar machine produced no bad output.
- A co-worker, Art Mirin, had also been experiencing similar problems with his CAM runs.



Node 587 effects – Last 10 years, 12-yr CAM run

Temperature Difference (Good - Bad)
(last 10 years of 12-year simulation)



(CAM/UQ MOAT3 study, run0035, monthly average history files)

**Regional Max Temperature Difference of
~1°C between Good & Bad runs.**

“Atlas Node 587” issue – The cause & remedy.

- Our systems people ended up finding an intermittently occurring small floating point roundoff error on one core of Node 587:
 - Order of one off in the least significant digit.
 - These roundoff errors propagated to significant differences in CAM diagnostic values over time.
- Checked all other nodes on Atlas & similar LLNL machines.
 - Found one more on another machine.
- Permanently removed both nodes.
 - Nodes sent to AMD for further analysis.

```
-2.181631512210958e+18  
-2.181631512210958e+18  
-2.181631512210959e+18  
-2.181631512210959e+18  
-2.181631512210959e+18  
-2.181631512210958e+18  
-2.181631512210959e+18  
-2.181631512210958e+18
```



“Atlas Node 587” issue – The fallout for us.

- Only made test runs for several weeks.
- Reran all runs that had used Node 587 – roughly 100.
- Used ~1M CPU-hrs?
- Now for every UQ run, we actually do two, one for 2 months & one for the desired number of years, & compare their outputs.
 - If they do not match, we know there is a problem.
 - The fact that they do match is a good thing, but does not guarantee that there is not a problem.
 - The cost of the extra 2 month runs is “tolerable”.
 - This would have caught the “Atlas Node 587” issue.



“Atlas Node 587” issue – Further investigation.

- Made some runs using different compiler optimization levels to look at CAM floating point precision issues in general (-O1/-O2).
- Found the level of variation induced by compiler optimization to be minor (~10%).
 - Similar level of variation to that seen in Node 587 runs.
- However, as ensemble results start to be filtered with observational data, these precision variations will become relatively more significant & will need to be reassessed.

