R&D for next generation meteorological simulations at RIKEN AICS

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Cartesian dynamical core, physical processes For regional weather/climate simulations Nishizawa et al (2015), Sato et al. (2015)



Icosahedral dynamical core

For global climate simulations

Tomita et al. (2001,2002), Tomita and Satoh (2004)

To be merged!







Easy comparison

Reproducibility

LES-scale simulation

WTK @ CPS

Reproducibility

Scientific products should be able to be replicated for verification and reliability.

Openness

• SCALE is available to anyone as an open source software.

Sharing know-how

- Predecessors' undocumented knowledges have often been lost.
- We try to publish knowledge of our experiences, e.g., parameter tuning, limiters...

Easy Comparison

Comparison is an important key in evaluation of the reliability of the meteorological numerical simulations.

Uncertainty of meteorological simulation

- not a first-principle simulation
- many empirical rules / hypotheses
- tones of tunable switches

Difficulty in validation of simulations

- limitation of observations (coverage, resolution, quantity)
- paleo/future climate, or other planets

Inter-model comparisons

total performance

Intra-model comparison

individual schemes

- physical processes, e.g.,
 - cloud microphysics: one/two moment bulk, spectral bin, super-droplet
- dynamical cores, e.g.,
 - discretization schemes
 - order of accuracy of difference scheme
 - implicit and explicit temporal integration schemes
- combination of the schemes
- tunable parameters
- precision of floating point

Differences are relatively easy to be understood

Comparison between cloud-microphysical schemes

RICO experiment (van Zanten et al. 2011)



We can conclude that these differences are originated from the cloud-microphysical schemes. <u>1-moment:</u> The faster drop is due to saturation adjustment and quick autoconversion.

Sato et al. 2015: Impacts of cloud microphysics on trade wind cumulus: which cloud microphysics processes contribute to the diversity in a large eddy simulation? PEPS, 2:23.

LES-scale simulations

Several added values are expected in high-resolution large-eddy simulations.

Smaller uncertainty, On more physical principles

- cumulus parameterization -> cloud microphysics
- RANS -> LES

Better representation of extremes

- finer topography / surface conditions
- less spatial averaging



• better data handling in pre/post processes

Validation of large grid aspect ratio (dx/dz) in LES

Unstable PBL turbulence experiment



<u>conventional SGS model</u>: spurious energy pile due to small mixing length

Nishizawa et al. 2015: Influence of grid aspect ratio on planetary boundary layer turbulence in large-eddy simulations, GMD, 8, 3393-3419.

Challenge to meso-scale LES

Huge domain with high resolution LES

- 300 km x 30 km domain with $\Delta x=50$ m, 275 layers
 - 1 billion grids
- 16 h integration (dt= 0.01 sec)
 - 138 h with 221,184 cores @K computer
- total 120TB output

Transition from closed to open cell of the stratocumulus



<u>Cloud cover</u> determined by the balance between distance of each cumulus and cloud broadening

Sato et al. 2015: Horizontal distance of each cumulus and cloud broadening distance determine cloud cover, SOLA, 11, 75-79.

Other planets

Highest resolution on Martian PBL experiment

- 19.8 km² domain with $\Delta x=5$ m, 3,300 layers
 - 50 billion grids
- 1 h integration (dt= 0.006 sec)
 - 200 h with 57,600 cores @K computer
- total 60TB output

Statistics of Martian dust devils



Nishizawa et al., 2016: Martian dust devil statistics from high-resolution large-eddy simulations, GRL.

SCALE Computational performance



performance @ K computer

- above 10% of peak performance (dynamical core)
 - 5~8% for full simulation (including I/O)
- about 100% weak scaling up to full system (663,552 cores)
- good strong scaling

Weak scaling





A Cost-effective Online Nesting Procedure

Conventional	onventional CONeP		domain 1	domain 2	domain 3
		grid space	27 km	9 km	3 km
		#grid (nx, ny, nz)	80, 80, 48	80, 80, 64	80, 80, 80
PD: time integ. to T1	INTERCOMM	time interval	27 s	9 s	3 s
by N nodes CD: time by N nodes PD: time integ. to T2 by N nodes CD: time to T2	by Mnodes by K nodes	time steps	50	150	450
	CD: time to T1	CNV	domain 1	domain 2	domain 3
	by Mnodes by K nodes	#node (mx, my)	10, 10	10, 10	10, 10
	PD: integ, to T2 CD: time byMnodes	#grid /tile (lx, ly)	4,4	6,6	10, 10
by I nodes		CONeP	domain 1	domain 2	domain 3
PD: integ. to T3	PD: to T3 CD: to T3	#node (mx, my)	2, 2	4,4	8, 10
CD: integ. to T3		#grid/tile (lx, ly)	40, 40	20, 20	10, 8
	continue continue				

renormance Experiment on R computer for 1350s time integration									
	Three domains	CNV	CONeP		Four domains	CNV	CONeP		
	elapsed time	20.9 s	16.8 s		elapsed time	61.3 s	44.9 s		
20% faster!			er!			27% faste	r!		

Yoshida et al., 2017: CONeP: A cost-effective online nesting procedure for regional atmospheric models, Parallel Computing.



Challenge! (explicit expression of cloud)

Our research community (NICAM research community)' approach: Resolve the cloud system & related process over the globe

NICAM development : ~2000

still development is continuing!

Conceptual development philosophy

• Explicit resolving the cloud itself

- Use of Icosahedral grid
 - To get a quasi-homogeneous grid for computational efficiency
- Nonhydrostatic DC
 - To resolve cloud scale (deep convection, shallow cloud etc.)
- Sophistication of cloud expression:
 - To avoid the ambiguity of cumulus parameterization and understand the cloud dynamics





Sub-km global simulation!

- Δx=870m, 94 layers
 - 63 billion grids
- 48 h integration (dt=2 sec)
 - 220 h with 163,840 cores @K computer
- total 320TB output
- 200-day for the post process on Xeon cluster
 ⇒ analysis on the K (163,840 cores)

A snapshot of sub-km AGCM









Convergence of convections with resolution

- Global composite of deep convection (vertical velocity)
 - Δx<2km: convection is represented at multiple grids



Efficiency of NICAM on K Computer

Performance efficiency

- Just after porting from ES : ~4%
- Cache optimization to stencil operators : ~5%
- Cleaning the time-wasting codes : ~7%
- Modify conditional branches, refactoring : ~10%



0.9PetaFLOPS

Weak scaling test

- Same problem size per node, same steps
- Good scalability

14km 3.5km 800m 600 400m 500 10 performance efficinecy[%] Elapsed time[sec] 400 8 300 6 200 4 100 2 0 0 5 20 80 320 1280 5120 20480 81920 23 WTK @ CPS number of nodes

NICAM870m/L96 animation

NICAM 870 m - 96 levels Real Case Simulation: 25 - 26, Aug., 2012

SPIRE field-3: Study of extended-range predictability using GCSRAM RIKEN / AICS: Computational Climate Science Research Team





Direction of our development in next 5 years

Infrastructure:

• Extension of basic library SCALE :

- <u>Corroborations w/ developers of other models</u>
 - Unified API to share components and save human resources
- Massive parallel analysis routines for acceleration of scientific output, social outcome
 - <u>Not only acceleration of simulation itself but also acceleration of analysis phase</u>

• Easy programing and high performance computing:

- DSL(Domain Specific Language) e.g. stencil DSL
 - w/ the Japanese next flagship computer project

Direction of our research in AICS in next 5 years

• Science:

BIG DATA assimilation:

- Now, developing....
 - NICAM + LETKF (with DA research team & post K priority subject)
 - Many satellite data is available.
 - One goal : Reanalysis data by cloud resolving model
 - SCALE+LETKF(with DA research team)
 - PA data provides tremendous information in time and space.
 - We are tackling to each cumulus with 30min lead time

<u>Reginal Climate assessment : downscale to city level</u>

- Disaster prevention and mitigation, adaptation
 - Multi-model ensemble (SCALE can do it!) drastically reduce the uncertainties for the future climate assessment in the regional model
 - Model bias reduction by data assimilation
 - e.g. Determination of unknown parameters

Planetary science

- Generalization of earth knowledge
- <u>Theoretical issue</u>
 - Moist LES theory



Dynamics

- Governing equations: 3-dimensional fully compressible
- Grid system : Arakawa-C type
- <u>Temporal integration:</u> HEVE, HEVI, HIVI
- Temporal difference: 3 steps Runge-Kutta scheme
- Spatial difference: 4th order central difference
- Topography: **Terrain-following**
- Positive definitive : FCT scheme

Other

- Offline/Online nesting system
- LETKE/assimilation system

Brief description of SCALE

Physical schemes

Cloud microphysics: Kessler (Kessler, 1969) 1-moment bulk (Tomita et al., 2008) 2-moment bulk (Seiki and Nakajima, 2014) 1-moment bin (Suzuki et al., 2010) super droplet method (Shima et al., 2009, experimental)



- Cumulus parameterization : Kain-Fritsch (*in preparation*)
- Radiation: ٠ MSTRN-X (Sekiguchi and Nakajima, 2008)
- Aerosol microphysics: 3-moment bulk (Kajino et al., 2013, experimental)
 - Surface flux: Louis-type (Uno et al. 1995) Beljaars-type (Beljaars and Holtslag 1994, Wilson 2001)
- Land: • Slab model with a bucket model
- Ocean: Slab ocean model
- Urban:

Single-layer urban canopy model (Kusaka et al., 2001)



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Convergence of 1. number of convection 2 distance of neighboring convection







Validation of higher resolution simulation

Density current test case

51.2 km x 6.4 km (2-D domain)

Same setting as Straka et al. (1993) but no physical diffusion.





back___

Revolutionary super-rapid data assimilation





Local Ensemble Transform Kalman Filter *(Hunt et al. 2007)*

Pinpoint (100-m resol.) forecast of severe local weather by updating 30-min forecast every 30 sec!

collaborate w/ AICS data assimilation Team, JMA, NICT, and Osaka Univ.

<u>Miyoshi et al. : "Big data assimilation" Revolutionizing severe weather prediction, BAMS,</u> accepted.

BDA 30-sec. assimilation cycle system

