

# Simulating Turbulence: Resolution, Extreme Events and Towards Extreme Computing

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*Turbulence from Angstroms to Light Years*

*Discussion Meeting in honor of Prof. Katepalli R. Sreenivasan*

International Center for Theoretical Sciences

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**Sreeni as a visionary, and catalyst**



## KRS70@Denver, Nov 17-18, 2017

# Simulation Database: a brief overview

“Deep Science” philosophy, to advance fundamental understanding

- Intermittency, inertial range and dissipative range
- Turbulent mixing, Schmidt numbers  $O(0.001)$  to  $O(1000)$
- Turbulent dispersion, backward tracking (w/ B.L. Sawford)

Forced isotropic turbulence,  $R_\lambda$  up to 1300; various resolutions

- Largest production run at  $8192^3$ , using 262,144 parallel processes
- Some shorter (but quite arduous) runs at  $12288^3$  (D. Buaria, poster at this conference) and  $16384^3$  (X.M. Zhai, M.P. Clay)
- Some 3 Petabytes stored at supercomputer centers

First paper with Sreeni published in 2002. His tremendous and benevolent influence has endured since mid 1990's. Visited over 10 times.

# Outline of This Talk

- Assessing the accuracy of large simulations
- Resolution in space and time for “extreme events”
- Towards extreme-scale computing: GPU algorithms



# As ever-larger simulations become possible...

Within the realm of 3D homogeneous turbulence:

- What do we want to use the computing power for?  
(higher Reynolds no., Mach no., other physical processes..)
- How do we recognize the limitations?  
(deterministic as well as statistical sources of errors...)

Can a simulation at higher resolution resolve any doubts?

- Higher moments, extreme fluctuations inherently more sensitive
- An expensive proposition, subject to statistical sampling
- Resolution in space (especially critical for work on intermittency)
- Resolution in time (seems to be discussed less in literature)

# Spatial Resolution in Pseudo-Spectral DNS

$$k_{max} = \sqrt{2}N/3; \quad \Delta x/\eta \approx 2.96/(k_{max}\eta)$$

- $k_{max}\eta \approx 1.5$  may be fine for low-order velocity statistics
- But better resolution needed for small-scale statistics, increasingly so at higher Reynolds numbers (Yakhot & Sreenivasan 2005).
- Refine the grid, run again at same  $Re$ , and compare (e.g. acceleration statistics, Yeung *et al.* PoF 2006)
- For a given snapshot, what features may be subject to errors?
  - ▶ Take best-resolved velocity field, truncate at some wavenumber  $k_c$ . Compute various statistics again, compare, and repeat
  - ▶ Large differences would indicate insufficient accuracy
  - ▶ A post-processing task, not a large new simulation
  - ▶ No information on global error after N-S time evolution

# Temporal Resolution

- Courant number constraint for numerical stability:

$$C = \left[ \frac{|u|\Delta t}{\Delta x} + \frac{|v|\Delta t}{\Delta y} + \frac{|w|\Delta t}{\Delta z} \right]_{\max} \sim \alpha \frac{u'\Delta t}{\Delta x}$$

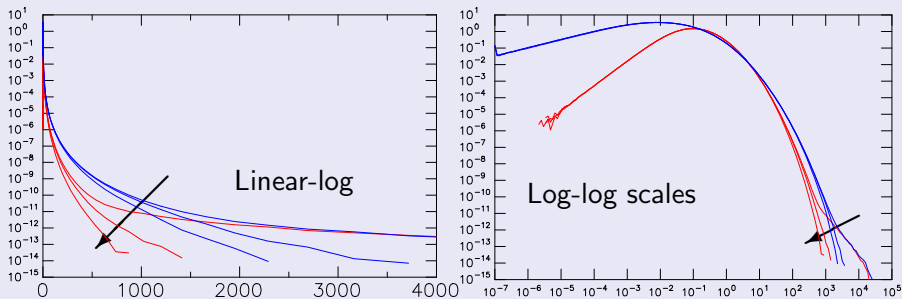
- Classical scaling and our experience in the simulations (take  $\alpha = 12$ )

$$\Delta t / \tau_\eta \approx (C/12)(15)^{1/4}(\Delta x / \eta) R_\lambda^{-1/2}$$

- In large simulations,  $\Delta t / \tau_\eta$  frequently well under 1%
- But small scales evolve fast (at fixed position, swept by large scale motions of speed  $\sim u'$ , with a time scale of order  $\eta / u'$ )
- Temporal intermittency is stronger than spatial intermittency, as comparisons between Eulerian and Lagrangian statistics suggest
- Short tests (approx 10  $\tau_\eta$ , say) at  $C = 0.6, 0.3, 0.15$  can help.

# Dissipation and Enstrophy: Spatial Filtering

- Enstrophy is more intermittent. Is it more sensitive to resolution?
- PDFs of  $\epsilon/\langle\epsilon\rangle$  and  $\Omega/\langle\Omega\rangle$ , derived from velocity fields at  $R_\lambda$  650,  $k_{max}\eta = 2.8$  ( $4096^3$ ); apply cutoff at  $k_c/k_{max} = 1, 0.75, 0.5$

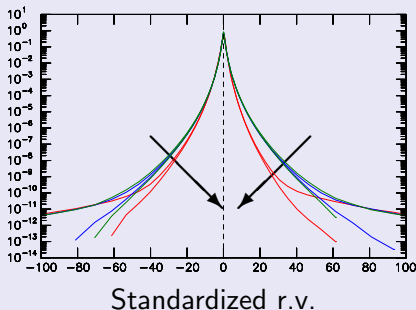


Convergence between PDFs of  $\epsilon$  and  $\Omega$  at far (power-law like) tails appears to be due to high- $k$  modes, contaminated by aliasing errors

# Why is dissipation more sensitive to resolution?

$$\epsilon \equiv 2\nu s_{ij}s_{ij} ; \quad \Omega \equiv \omega_i\omega_i$$

- For incompressible isotropic turbulence:  $\nabla_{\parallel}\mathbf{u}$  is negatively skewed and contributes only to  $\epsilon$ ; while  $\nabla_{\perp}\mathbf{u}$  is more intermittent



- Narrower PDF tails when some high  $k$  modes removed by filtering.
- $\nabla_{\parallel}\mathbf{u}$  more sensitive to resolution than  $\nabla_{\perp}$  or  $\nabla \times \mathbf{u}$

# Incompressibility and 1-D Spectra

Incompressibility:  $\hat{\mathbf{u}}(\mathbf{k}) \perp \mathbf{k}$  in wavenumber space

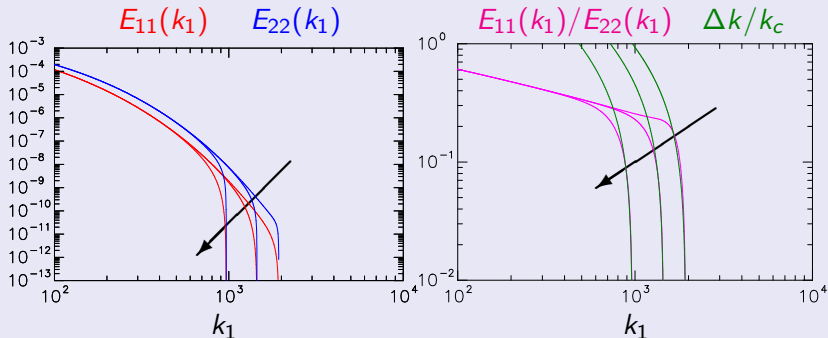
- At high  $k \equiv |\mathbf{k}|$ , spectral content of  $\hat{u}_1$  is mostly in Fourier modes with large  $k_2$  and/or  $k_3$  (i.e.,  $k_\perp$ )

Relations for 1D spectra with  $k_1$  near cutoff  $k_c$  ( $k \approx k_1$ )

$$\begin{aligned} E_{11}(k_1) &= \int_{k_1}^{\infty} \frac{E(k)}{k} (1 - k_1^2/k^2) dk \\ &\approx \frac{E(k_1)}{k_1} \int_{k_1}^{k_c} (1 - k_1^2/k^2) dk \approx E(k_1)(\Delta k/k_1)^2 \\ E_{22}(k_1) &= \frac{1}{2} \int_{k_1}^{\infty} \frac{E(k)}{k} (1 + k_1^2/k^2) dk \approx E(k_1)(\Delta k/k_1) \end{aligned}$$

# Incompressibility and 1-D Spectra (cont'd)

- A spherical cutoff filter based on  $k$  removes more of high wavenumber content in longitudinal spectra than transverse (hence affects  $\nabla_{\parallel} \mathbf{u}$  more than  $\nabla_{\perp} \mathbf{u}$ )
- For small  $\Delta k = k_c - k_1$ :  $E_{11}(k_1)/E_{22}(k_1) \approx \Delta k/k_c$
- $R_{\lambda} \sim 650$ ,  $k_{max}\eta \approx 2.8$ ,  $4096^3$ ,  $k_c/k_{max} = 0.75, 0.5$ :



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# Tests of Spatial and Temporal Resolution

A wide parameter space: easier if not at the highest Reynolds number:

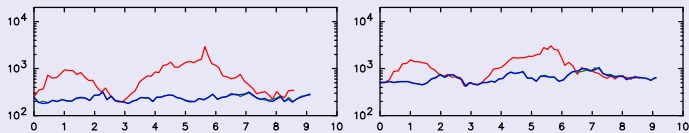
$R_\lambda \sim 390$				
$k_{max}\eta$	$N$	$C = 0.6$	$C = 0.3$	$C = 0.15$
1.33	1024	1.31	0.699	0.349
2.67	2048	0.609	0.333	0.165
5.38	4096	0.327	0.164	0.082
$R_\lambda \sim 650$				
1.33	2048	0.646	0.484	0.238
2.67	4096	0.411	0.236	0.116
5.38	8192	0.205	0.102	0.062

Resolution parameters and [percentage](#)  $\Delta t/\tau_\eta$  [Note: RK2 in time] for simulation datasets (roughly 10  $\tau_\eta$ ) at Reynolds numbers corresponding (within statistical error) to past simulations at  $R_\lambda$  390 and 650.

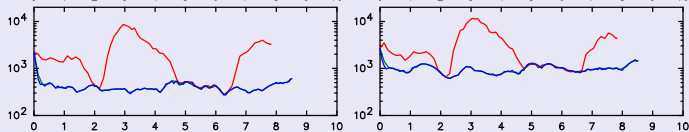
# Resolution in **time**: Peak dissipation and enstrophy

$R_\lambda \sim 390$ ,  $C = 0.6, 0.3, 0.15$ ;  $\epsilon/\langle\epsilon\rangle$  and  $\Omega/\langle\Omega\rangle$  vs.  $t/\tau_\eta$

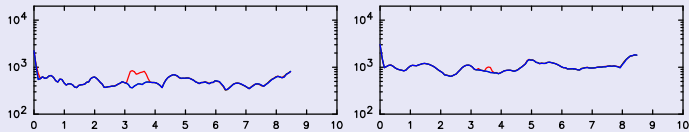
$k_{max}\eta \approx 1.3$



$k_{max}\eta \approx 2.7$



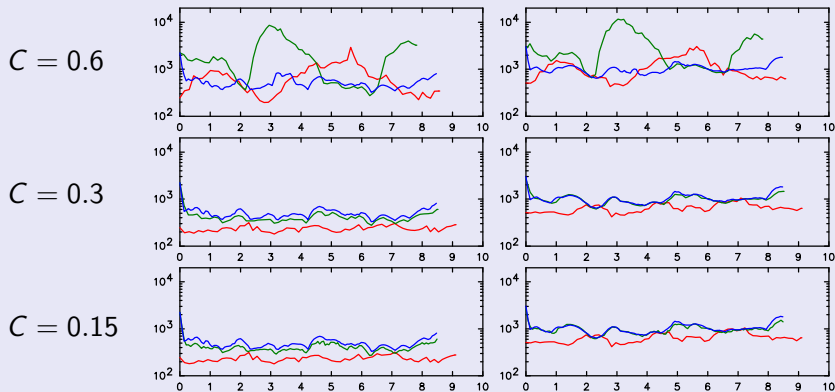
$k_{max}\eta \approx 5.4$



- $C = 0.6$  gives spuriously large peaks; 0.3 and 0.15 almost the same
- Sensitivity greater for  $\epsilon/\langle\epsilon\rangle$ , weaker at higher  $k_{max}\eta$  (such as 5.4)

# Resolution in **space**: Peak dissipation and enstrophy

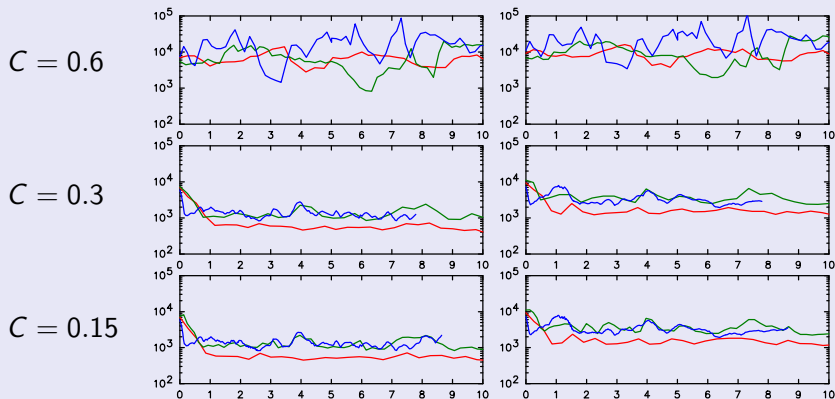
$R_\lambda \sim 390$ ,  $k_{max}\eta = 1.3, 2.7, 5.4$ ;  $\epsilon/\langle\epsilon\rangle$  and  $\Omega/\langle\Omega\rangle$  vs.  $t/\tau_\eta$



- Impact of using higher  $k_{max}\eta$  not clear if  $C$  is too high
- Low  $C$ : higher  $k_{max}\eta$  does allow larger gradients,  $\epsilon$  and  $\Omega$

# Would a lower Courant no. be needed at higher $R_\lambda$ ?

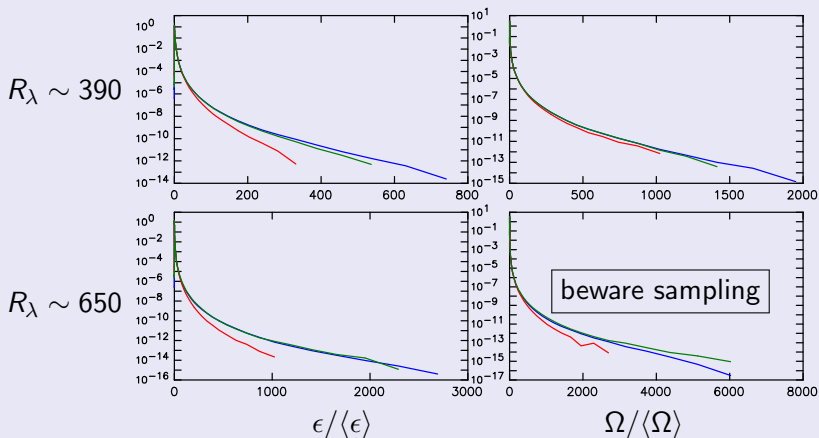
$R_\lambda \sim 650$ ,  $k_{max}\eta = 1.3, 2.7, 5.4$ ;  $\epsilon/\langle\epsilon\rangle$  and  $\Omega/\langle\Omega\rangle$  vs.  $t/\tau_\eta$



Not perfectly clear, but sensitivity to  $C$  at  $k_{max}\eta \sim 5.4$  seems stronger than at  $R_\lambda \sim 390$ , even as  $\Delta t/\tau_\eta$  drops

# Best results available for dissipation and enstrophy PDFs

$C = 0.15$ ,  $k_{max}\eta \approx 1.3 \ 2.7 \ 5.4$ : Convergence apparently achieved



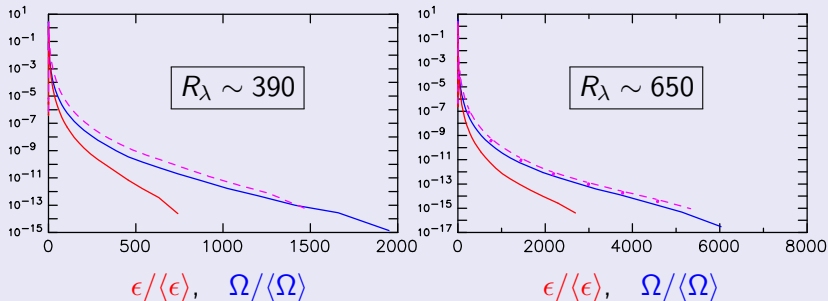
Tails stretch out further: as  $R_\lambda$  increases, and  $\Omega$  relative to  $\epsilon$

# Re-visiting the issue of dissipation vs enstrophy PDFs

- Both are measures of the small scales: should they scale similarly at sufficiently high Reynolds no.? (Nelkin 1999)
- Most data sources, including DNS at finite Reynolds no., show that enstrophy is more intermittent
- Visualizations of spatial structure: vortex filaments in contrast to more diffuse, sheet-like dissipative structures
- Recent DNS at higher resolution suggest both  $\epsilon$  and  $\Omega$  attain amplitudes more extreme than previously reported, scaling similarly unless averaged over significant scale sizes (Yeung *et al.* 2015)
  - ▶ yet contaminated by the effect of time-stepping errors, which affect dissipation more than enstrophy
  - ▶ although  $\Delta t \ll \tau_\eta$ , need to have  $\Delta t \ll \eta/u'$ , to resolve rapid rate of small scales advected by the large scales (Tennekes JFM 1975)

# Compare dissipation and enstrophy PDFs again

Best data at  $C = 0.15$  (RK2),  $k_{max}\eta \approx 5.4$ : tails do not coincide, but seem to have the same shapes (Magenta dotted line is PDF of  $2\epsilon/\langle\epsilon\rangle$ )



Possible explanation from multi-fractal theory

- Assume  $\langle\epsilon^n\rangle \propto R_\lambda^{f(n)}$ ,  $\langle\Omega^n\rangle \propto R_\lambda^{g(n)}$ , for  $n \geq 1$ . [with  $f(1) = g(1)$ ]
- If  $f(n)$  and  $g(n)$  have the same functional forms: at a given  $R_\lambda$ ,  $\exists \beta$  s.t.  $\beta\epsilon/\langle\epsilon\rangle$  and  $\Omega/\langle\Omega\rangle$  have nearly the same PDF at large amplitudes

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# How will we do larger simulations?

How do we do it now...

- Massive distributed-memory parallelism, with 2D domain decomposition (3D domain as “pencils”)
- Usually, all-to-all communication for transposes of data is dominant
- On “Blue Waters” (Cray XE6): remote memory addressing via Co-Array Fortran, also helped by favorable machine topology

Changes in HPC landscape, towards Exascale (by 2021 in US)

- Many-core platforms with fewer but fatter nodes
- Heterogeneous computing: especially multi-threaded GPUs for fast arithmetic, but data movements are (even more) the key

GPUs: recent success for turbulent mixing at high Schmidt number using a combination of FFTs and compact finite differences (Gotoh *et al.* 2012)

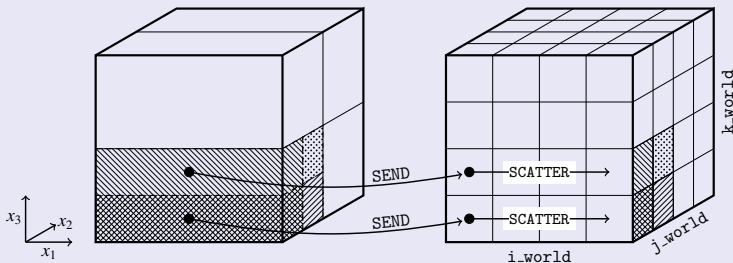
# Turbulent mixing at high Schmidt number

- $Sc = \nu/D$  varies: 7 for heat in water, 700 for salinity in ocean
  - $Sc \gg 1$ : smallest scale is Batchelor scale  $\eta_B = \eta/\sqrt{Sc}$ . Different scaling laws, and very difficult to resolve
  - Disparate resolution requirements:
    - ▶ velocity on coarser grid, scalar on finer grid
    - ▶ scalar: use numerical method with lesser communication needs
    - ▶ some overlap of operations on velocity and scalar field?
    - ▶ CPU host to work on velocity, GPU for scalar
  - Dual grid, dual communicator, dual scheme algorithm
    - ▶ CPU-based: Clay *et al.* 2017 (*Comput. Phys. Commun.*, published)
    - ▶ GPU-based: Clay *et al.* 2018 (*Comput. Phys. Commun.*, in revision)
- GPU: Performed well on 27-Petaflop TITAN (Cray XK7) at Oak Ridge

# Dual-Communicator Algorithm

Clay et al. *Comput. Phys. Commun.* 2017

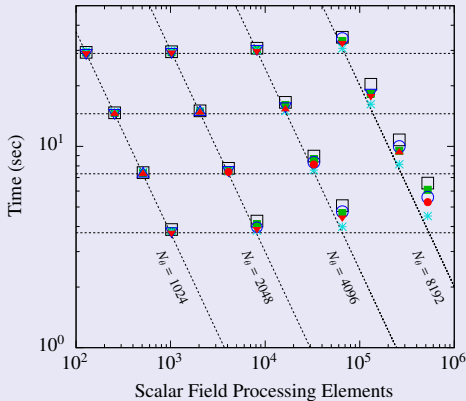
- Say  $N_\theta/N_u = 8$ : workload on finer grid is much heavier
- Define 2 groups of MPI processes, proportional to  $N_u^3$  and  $N_\theta^3$
- One-way intercommunicator transfer: need **u** on scalar side



- Overlap velocity field computations with scalar
- Overlap between operations on different velocity components

# CPUs: DNS Code Performance Comparisons

- Problem sizes  $1024^3$  to  $8192^3$ , different implementations
  - ▶ black squares: blocking and single-threaded
  - ▶ cyan stars: overlapping, using OpenMP nested parallelism



- Also very high scalability
- About 65% of time is spent on CCD3D
- Interpolation approx 25%
- Performance gain less than for CCD3D, due to reduced overlapping

# GPU Programming Considerations

[Mainly based on experiences on TITAN (Cray XK7, 1 GPU/Node)]

- Directives-based, instead of CUDA, use OpenACC or (better yet) OpenMP 4.5 standard (facilitates asynchronous execution)
- FFTs using CUDA mathematical libraries extremely fast
- Memory of GPU only 1/5 of host: needs careful management
- Minimize data movement: almost all scalar field operations on GPUs
- Keep GPU(s) as busy as possible: e.g. by using 2 MPI processes on 16-core TITAN node sharing 1 GPU
- NVIDIA visual profiler can provide detailed diagnostics
- Currently, all communication go through CPUs, thus extra copying between CPU and GPU. Hope to avoid this on future machines

# GPU: Weak Scaling and GPU-to-CPU Speedup

$N_\theta^3$	512 <sup>3</sup>	1024 <sup>3</sup>	2048 <sup>3</sup>	4096 <sup>3</sup>	8192 <sup>3</sup>
Nodes	2	16	128	1024	8192
CPU (s)	14.80	14.94	15.33	15.71	16.13
Weak (%)	—	99.1	96.5	94.2	91.8
Async. Off	2.943	3.042	3.155	3.441	3.686
Weak (%)	—	96.7	93.3	85.5	79.8
Speedup	5.03	4.91	4.86	4.57	4.38
Async. On	2.929	2.900	2.938	3.024	3.259
Weak (%)	—	101	99.7	96.7	89.9
Speedup	5.05	5.15	5.22	5.20	4.95

# Conclusions and Next Steps: Science Aspects

Resolution effects and numerical challenges in time and space stronger than anticipated earlier (at RK2, needs Courant no.  $C = 0.3$  or  $0.15$ )

- Although less intermittent than enstrophy, dissipation more sensitive to the numerics (role of incompressibility, vortex filaments, etc)
- Some past statements about extreme events in need of revision
- To examine/quantify resolution effects on various other statistics (perhaps inertial range statistics only weakly affected)

Successful development and execution of highly optimized algorithm for mixing at high  $Sc$  on a 27-PF GPU machine

- Minimizing data movement and aggressive overlapping using OpenMP 4.5 on GPUs: factor of 5 speedup over CPU only
- To fully analyze passive scalar with mean gradient at  $R_\lambda \sim 140$ ,  $Sc = 512$ ; and extend to active scalars (stratified flow)

# Conclusions and Next Steps: Computing Aspects

Viability of  $16384^3$  or higher on a **production** basis

- Adopt OpenMP 4.5 ideas in recent work on mixing at high  $Sc$ , towards new pseudo-spectral algorithm for velocity field
- Port codes to next big machine (150 PF SUMMIT at ORNL)
- Extension from 1 scalar to 2; passive to active

Obtaining, using, preserving resources

- Make science case in competitive proposals
- Large machines: problems are to be expected
- Where to put data and even hold on to them long term

More community and international cooperation would be very useful



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- Dhawal Buaria (PhD 2016), Max Planck Inst, Germany
- Matthew P. Clay (PhD 2017), Air Force Research Lab, USA

## Current PhD Students

- Shine X.M. Zhai, Kiran Ravikumar

# Some Publications

- Buaria, D., PKY, & Sawford, B.L. (2016) A Lagrangian study of turbulent mixing: forward and backward dispersion of molecular trajectories in isotropic turbulence. *JFM* **799**, 352-382.
- Clay, M.P., Buaria, D., Gotoh, T. & PKY (2017) A dual communicator dual grid-resolution algorithm for Petascale simulations of turbulent mixing at high Schmidt number. *Comput. Phys. Commun.*, **219**, 313-328.
- Donzis, D.A., Sreeni & PKY (2010) The Batchelor spectrum for mixing of passive scalars in isotropic turbulence. *Flow, Turb. & Combust.*, **85**, 549-566.
- Donzis, D.A., PKY & Sreeni (2008). Energy dissipation rate and enstrophy in isotropic turbulence: resolution effects and scaling in direct numerical simulations. *PoF*, **20**, 045108.
- PKY, Donzis, D.A. & Sreeni (2012). Dissipation, enstrophy and pressure statistics in turbulence simulations at high Reynolds numbers. *JFM*, **700**, 5-15.
- PKY, Zhai, X.M & Sreeni (2015). Extreme events in computational turbulence. *PNAS*, **112**, 12633-12638.