

Direct Numerical Simulation of Turbulent Flows

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Is turbulence a solved problem?

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Engineering: conceptually yes

- (depends on the complexity of your problem)
- relevant scenarios are computationally accessible



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- statistical description from first principles is missing
- → non-equilibrium thermodynamics?



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Mathematics: no

- boundedness of solutions with smooth initial conditions?



Introduction

Problem: turbulence ...

- is described by nonlinear equations
- exhibits spatio-temporal chaos
- involves large space- and time-scales



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Possible solutions:

- understanding of structures
- formulating a statistical theory



Introduction

Problem: turbulence ...

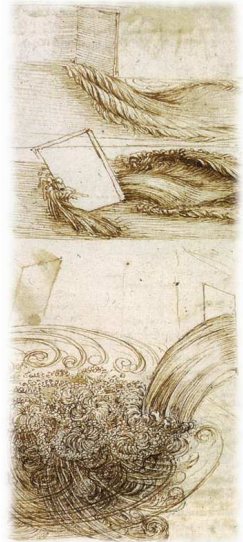
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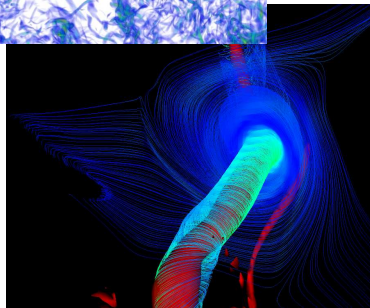
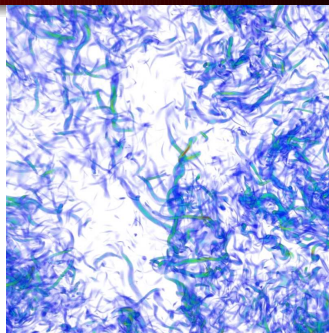
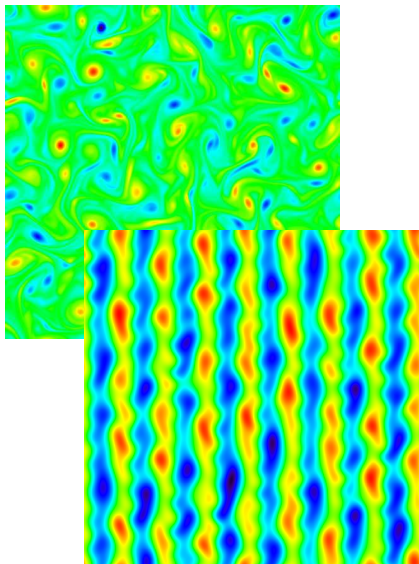
Tools:

- any kind of mathematics, that will do
- computer simulations



DNS

DNS: Visualization



DNS: equations

Navier-Stokes equations:

$$\frac{\partial \mathbf{u}}{\partial t}(\mathbf{x}, t) + \mathbf{u}(\mathbf{x}, t) \cdot \nabla \mathbf{u}(\mathbf{x}, t) = -\nabla p(\mathbf{x}, t) + \nu \Delta \mathbf{u}(\mathbf{x}, t) + \hat{\mathbf{f}}(\mathbf{x}, t)$$

$$\nabla \cdot \mathbf{u}(\mathbf{x}, t) = 0$$

DNS: equations

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Vorticity: $\boldsymbol{\omega}(\mathbf{x}, t) = \nabla \times \mathbf{u}(\mathbf{x}, t)$

Vorticity equation:

$$\frac{\partial \boldsymbol{\omega}}{\partial t}(\mathbf{x}, t) = \nabla \times (\mathbf{u}(\mathbf{x}, t) \times \boldsymbol{\omega}(\mathbf{x}, t)) + \nu \Delta \boldsymbol{\omega}(\mathbf{x}, t) + \mathbf{f}(\mathbf{x}, t)$$

DNS: numerics I

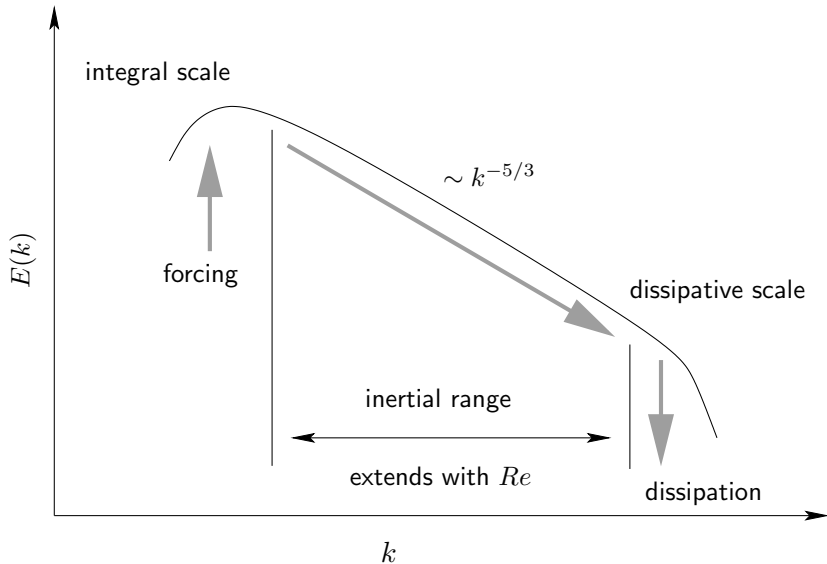
- aim: forced (stationary) homogeneous, isotropic turbulence
- temporal discretization: RK3 TVD
- spatial discretization: box-length 2π , dim grid points, periodic boundary conditions
- pseudospectral code

DNS: numerics II

$$\frac{\partial \tilde{\omega}}{\partial t}(\mathbf{k}, t) + \nu k^2 \tilde{\omega}(\mathbf{k}, t) = i\mathbf{k} \times \mathcal{F}\{\mathbf{u}(\mathbf{x}, t) \times \boldsymbol{\omega}(\mathbf{x}, t)\} + \tilde{f}(\mathbf{k}, t)$$

- adaptive time-stepping (Courant-Friedrichs-Levy criterion)
- *pseudospectral*: forward/backward FFT is computationally cheaper than convolution ($N \log N$ vs. N^2)
- aliasing: smooth Fourier filter
- viscosity is treated exactly (integrating factor)
- forcing: freezing of low modes
- code is MPI parallel

DNS: computational costs I



DNS: computational costs II

- forcing scale and dissipative scale should be well separated
- inertial range extends with increasing Re
- size of smallest structures decreases with Re
- smallest structures should be well-resolved by the grid
- turbulent field should be accurately advanced in time

DNS: computational costs III

to be more precisely ...

$$\eta = \left(\frac{uL}{\nu} \right)^{-3/4} L = Re^{-3/4} L$$

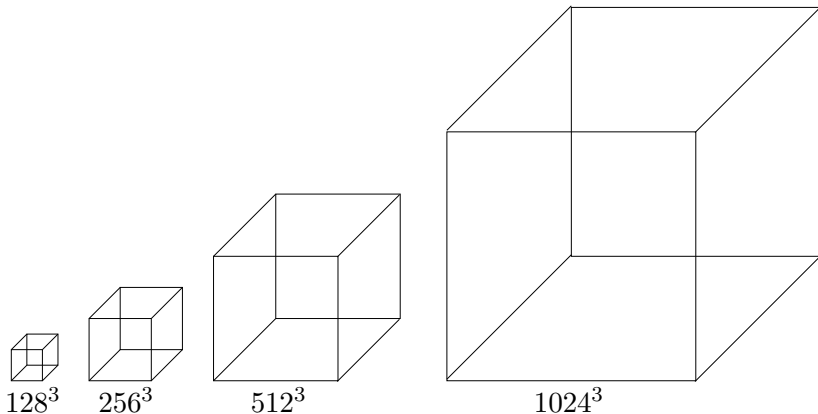
$$\Delta x \sim \eta$$

$$N_x \sim \left(\frac{2\pi}{\Delta x} \right)^3 \sim \left(\frac{2\pi}{L} \right)^3 Re^{9/4} \longrightarrow Re \sim \left(\frac{L}{2\pi} \right)^{4/3} N_x^{4/9}$$

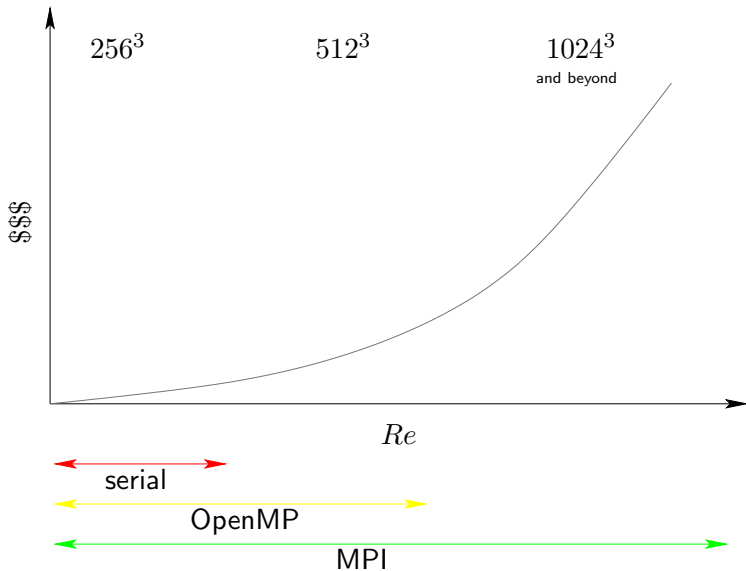
$$N_t \sim \frac{T}{\Delta t} \sim \frac{T}{\Delta x/u} \sim \frac{T}{l/u} Re^{3/4}$$

$$$$$ \sim N_x N_t \sim \left(\frac{T}{l/u} \right) \left(\frac{2\pi}{l} \right)^3 Re^3$$

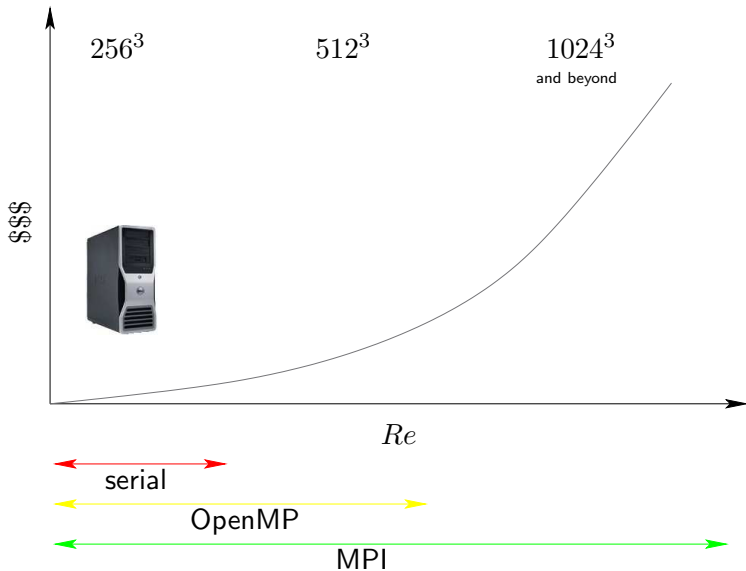
DNS: computational costs IV



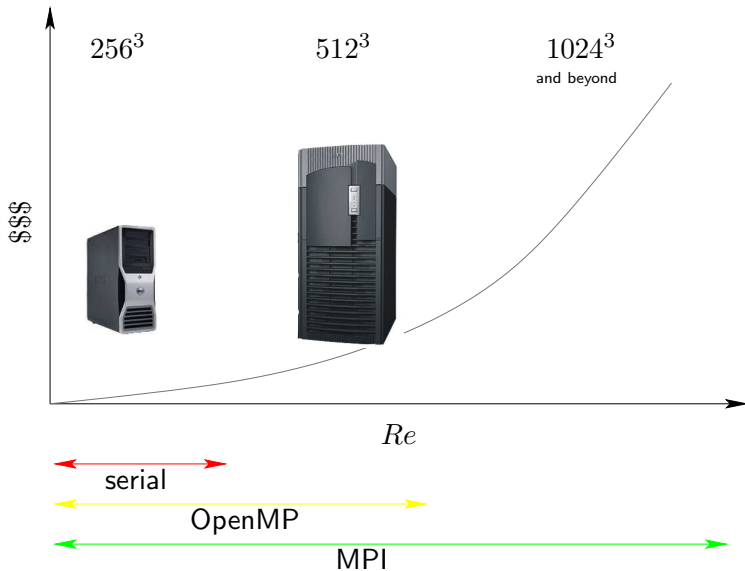
DNS: computational costs V



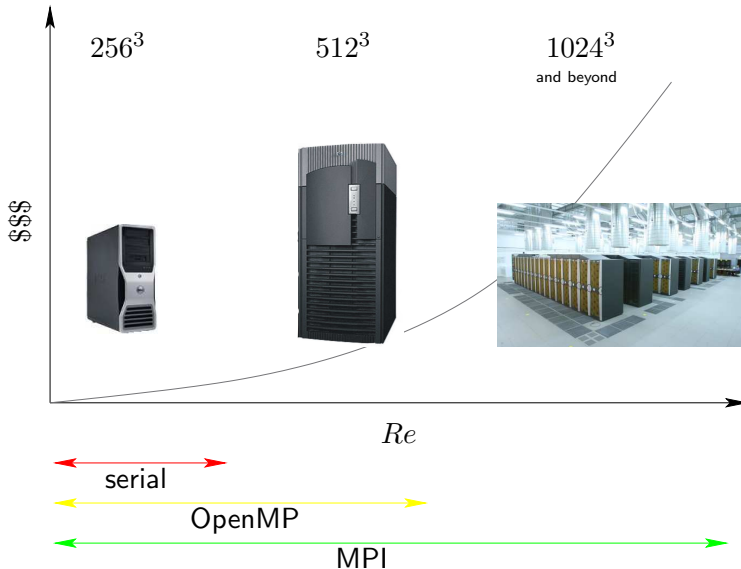
DNS: computational costs V



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DNS: computational costs V



DNS: OpenMP vs. MPI

OpenMP

- “fork and join” principle
- easy
- incremental parallelization
- medium number of cores
- limited degree of scalability

MPI

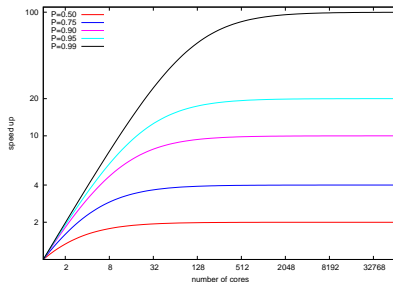
- decomposition of computational domain
- “communicate when necessary” principle
- broad range of application
- high scalability

Amdahl's Law

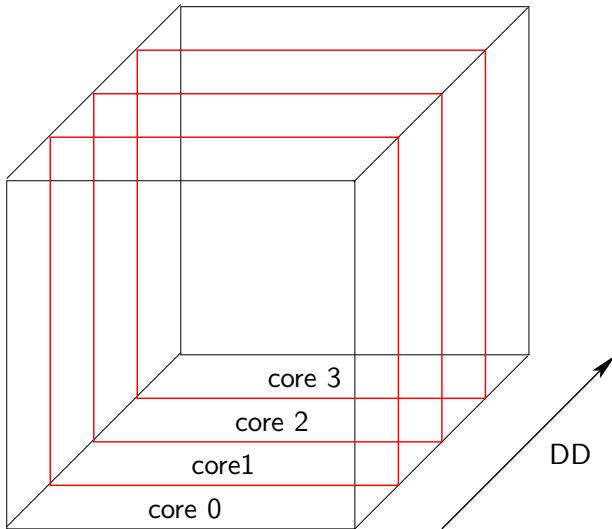
- consider serial code with runtime 1
- P denotes the parallelizable fraction of code
- runtime on n cores: $(1 - P) + \frac{P}{N}$
- maximum expectable speed up $S(N) = \frac{1}{(1-P) + \frac{P}{N}}$

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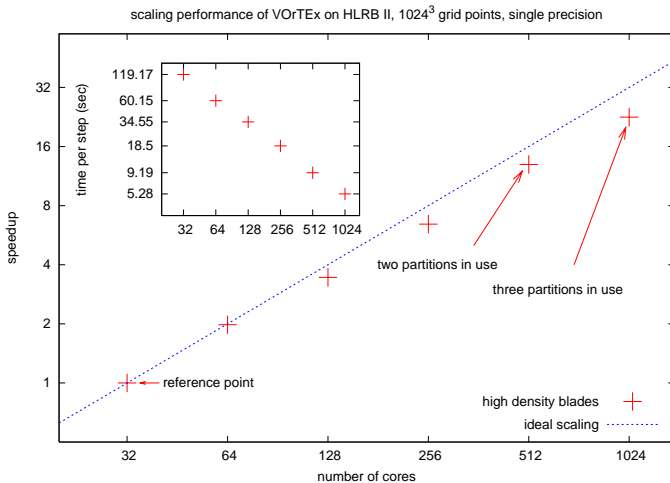
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DNS: parallelization via MPI



DNS: scaling



Demos

- Hello OpenMP/ MPI world
- HLRBII @ LRZ
- remote visualization