Towards Real-Time CFD Simulation of Indoor Environment

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Introduction

Indoor environment. State of art

Heating, ventilation and air conditioning (HVAC) systems control main indoor air parameters (temperature, velocity, relative humidity, etc.) and create comfortable indoor environment.

Air distribution can be typically evaluated by:

- analytical models
- experimental measurements
- computer simulations

The complexity of indoor airflow makes experimental or analytical investigation extremely difficult and expensive.



Introduction

Indoor environment simulation models

Multizone (airflow network) models - low computational cost and low accuracy.

Zonal models - moderate computational cost and moderate accuracy, but high case dependence.

Computational Fluid Dynamics (CFD) - high computational cost and high accuracy.



Introduction

Objectives

To investigate the capabilities of CFD to perform real-time and even faster simulations of indoor environment.

To choose a reliable model to perform CFD simulations of indoor environment with minimal computational cost and adequate accuracy.

To discuss possibilities of using CFD for design and short-term thermal behaviour prediction purposes.



Physical problem and governing equations

Problem description



¹D. Saury, N. Rouger, F. Djanna and F.Penot. Natural convection in an air-filled cavity: Experimental results at large Rayleigh numbers.*Int Comm Heat Mass Transf*, 38(6):679-687, 2011.

Physical problem and governing equations Governing equations

$$\nabla \cdot \mathbf{u} = 0$$

$$\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla)\mathbf{u} = PrRa^{-1/2}\nabla^2 \mathbf{u} - \nabla p + \mathbf{f}$$

$$\frac{\partial T}{\partial t} + (\mathbf{u} \cdot \nabla)T = Ra^{-1/2}\nabla^2 T$$

$$\mathbf{f} = (0, PrT, 0)$$

Reference quantities:

- $t_{ref} = Ra^{1/2}H^2\alpha^{-1}$
- $u_{ref} = Ra^{1/2}(\alpha/H)$
- $T_{ref} = \Delta T = T_h T_c$

RANS approach

RANS simulations are performed by the open source CFD code **OpenFOAM** using "*buoyantBoussinesqPimpleFoam*" solver.

RANS models used:

- **SST** $k \omega$ model under-predicting overall heat transfer, highest computational cost among all three models. comparison with LES models.
- $\mathbf{k} \epsilon$ model good overall performance, low computational cost.
- **RNG** $k \epsilon$ model good overall performance, computational cost is higher than $k \epsilon$ model.

LES discretization

Collocated grid - in-house CFD code **Termofluids** applied to unstructured meshes.

Staggered grid - in-house CFD code **STG** based on a fourth-order symmetry-preserving finite volume discretization on structured grids.

LES on collocated grid

Simulations are performed by the in-house CFD code Termofluids.

LES models used:

- LES-WALE good overall performance, moderate computational cost.
- LES-VMS(WALE) good overall performance, high computational cost.

LES on staggered grid

Simulations are performed by the in-house **STG** based on a fourth-order symmetry-preserving finite volume discretization on structured staggered grids.

LES models used:

- LES-WALE good overall performance, moderate computational cost.
- LES-QR least accurate results for the coarse grids, moderate computational cost.
- LES-S3PQ good overall performance, low computational cost.

Models chosen

Collocated grid:

- RANS $k \epsilon$
 - LES-WALE
 - No model

- Staggered grid:
 - LES-S3PQ
 - No model

Simulation details

Case	N _x	Ny	Nz	N _{total}	γ_{x}	$(\Delta x)_{min}$
M0 (DNS)	450	900	256	$1.04 imes10^{8}$	2.00	$4.28 imes10^{-5}$
M1	8	30	4	$9.60 imes10^2$	2.00	$7.97 imes10^{-3}$
M3	12	50	8	$4.80 imes10^3$	2.00	$4.46 imes10^{-3}$
M5	18	80	12	$1.73 imes10^4$	2.00	$2.70 imes10^{-3}$
M7	30	120	20	$7.20 imes10^4$	2.00	$1.49 imes10^{-3}$
M10	70	240	40	$6.72 imes10^5$	2.00	$5.40 imes10^{-4}$
M11	100	320	40	$1.28 imes10^{6}$	2.00	$4.05 imes10^{-4}$

Simulation duration: DNS - 600 time units; LES - 1200 time units; RANS - 250 time units.



Verification





Verification





Accessing real-time and faster than real-time simulations

Reference machine: AMD Opteron 2350, 24Gb/s memory bandwidth, 1 - 32 cores and 1-4 nodes.

Target machine: Intel Core i7-8700K, 41.6Gb/s memory bandwidth, 6 CPU cores and 1 node.

$$t_{tgt} = t_{ref} \frac{BW_{ref}}{BW_{tgt}} \frac{CPU_{ref}}{CPU_{tgt}} \frac{NODE_{ref}}{NODE_{tgt}}$$
$$R = \frac{t_{wc}}{t_{phy}}$$
$$R < 1$$

Global quantities checked

Average Nusselt number $< Nu >= \int_0^1 \frac{\partial < T >}{\partial x} dy \Big|_{x=0}$ Average stratification $< S >= \frac{\partial < T >}{\partial y} dx \Big|_{y=0.5}$

Average kinetic energy

Average enstrophy

$$\langle E \rangle = \int_V \frac{\langle \mathbf{u}^2 \rangle}{2} dV$$

$$<\Omega>=\int_V<\omega^2>dV$$

Global quantities - < Nu >



Global quantities - < S >



Global quantities - < E >



Global quantities - < E > zoomed



Global quantities - $< \Omega >$



Model chosen

LES-S3PQ model on staggered grid showed the best overall performance and is chosen for further tests.

Applications

Potential of real-time simulations for design and control purposes

	Conceptual design	Detailed design	Control
Simulation duration	24 hours	24 hours	15 minutes
Time ratio	$R \leqslant 0.5$	$R \leqslant 0.5$	$R \leqslant 0.1$
Computational resources	6 cores	72 cores	6 cores
Relative error	15 %	5 %	5 %

Applications

Potential of real-time simulations for design and control purposes

	Conceptual design	Detailed design	Control
Grid resolution	17280	672000	672000
Wall-clock time, sec	$t_{wc} = 1533$	$t_{wc} = 40690$	$t_{wc} = 122057$
Time ratio	$\textrm{R}\approx 1.3$	Rpprox 34	$\textrm{R}\approx 100$

Applications

Potential for the future applications



Accessing real building scale

Experimental scale:	Simulated scale:	Generic building scale:
$\Delta T = 20^{\circ}C$	$\Delta T = T_{hot} - T_{cold} = 28^{\circ}C - 23^{\circ}C = 5^{\circ}C$	$\Delta T = T_{hot} - T_{cold} = 28^{\circ}C - 23^{\circ}C = 5^{\circ}C$
$L \times H \times D =$ $1.00 \times 3.84 \times 0.86m$	L imes H imes D = 1.59 imes 6.10 imes 1.37m	$L \times H \times D = 10 \times 10 \times 10 m$
$V = 3.30 m^3$	$V = 13.29m^3$	$V = 1000 m^3$

Accessing real building scale

A generic building is approximately 75 times bigger than simulated cavity.

Assuming that the computational cost scales linearly with the volume.

 $R \approx 100$ for 15% relative error using 6 CPU cores. $R \approx 7650$ for 5% relative error using 6 CPU cores. $R \approx 2550$ for 5% relative error using 72 CPU cores.

Complete building scale

CFD simulations for a complete building would be available in less than 10 yeas for early design stage, in approximately 15 years for detailed design and within three decades for MPC systems.



Conclusions

- LES turbulence models in overall show better performance than RANS.
- The simulations stability could be improved using a staggered symmetry-preserving discretization together with turbulence modelling.
- At this moment CFD simulations are not affordable for indoor environment.
- CFD would be available for design purposes within two decades.
- HVAC predicting control systems equipped with CFD simulation tools will get affordable in approximately two-three decades.

Future work

- The problems analyzed would be extended to ventilated cavities with heat sources and domains with complex geometry.
- The simulations financial costs in terms of electrical power spent per CPU hour would be analyzed.
- Existing numerical algorithms would be optimized and accelerated using a fully-portable, algebra-based framework for heterogeneous computing on GPUs.

THANKS FOR YOUR ATTENTION!

Ready for your questions!