



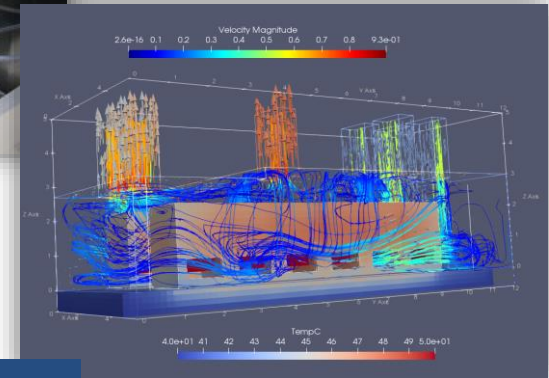
**Sensitivity studies OD-1D/3D of thermo
aeraulic models of industrial premises**

**Assistance in the identification of
influential phenomena**

**Preparation of experimental protocols
for the Zephyr test platform**

OpenTURNS Days – June 10th 2021

Pascal Borel (R&D/PRISME)

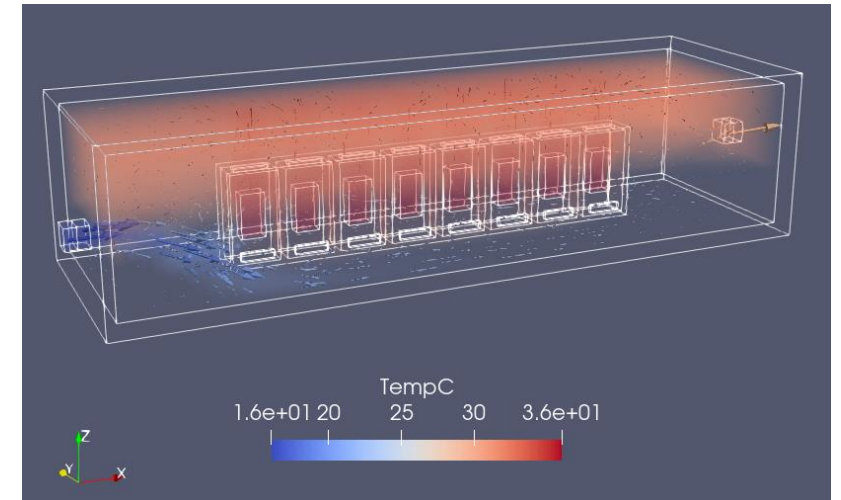
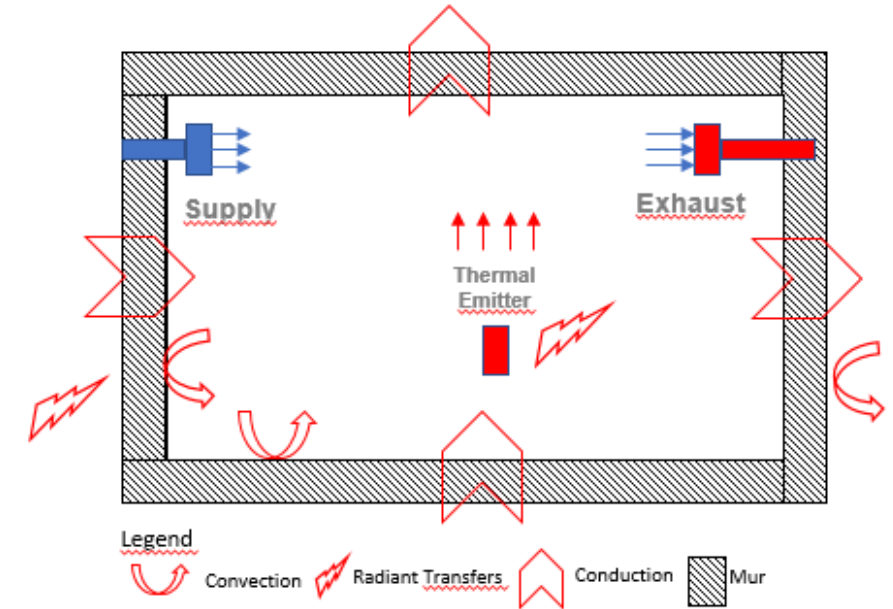


Proposed Agenda

1. **Context of industrial thermo-aeraulic studies**
2. **Setup of numerical study experiment of Zephyr laboratory**
 - a. Presentation of Zephyr laboratory
 - b. Numerical methodology
 - c. Salome Workflow
 - d. Variables of interest
 - e. Uncertain parameters
3. **Results and Perspective**
 - a. Experience plan results analysis: distribution and dependency
 - b. Metamodeling : sensitivity studies

Context of industrial thermo-aeraulic studies

- **Context – Objectives of industrial thermal studies**
 - Assess operability of electrical and I&C cabinet in industrials premises in various situation (consistency with thermal qualification)
 - Thermal transient : steady state, thermal transient with partial or total loss of HVAC
 - Premises configuration : geometry, thermal loads, thermal conditioning
- **Industrial practices**
 - Use of 0D/1D codes with a unique thermal potential for the whole cavity :
 - no access to heterogeneity indicators.
 - Some rooms can appear as critic with low margin between predicted temperature and materiel qualification temperature
- **Numerical Study objectives**
 - Preliminary identification of **influent phenomena** on **temperature distribution**
 - **Compare 0D/1D-3D** code predicted temperatures / thermal power balances (convective / radiative heat flow)
 - Focused on geometry of **Zephyr laboratory** test room (building in progress on EDF R&D Chatou site)



Modular laboratory ZEPHYR - Generalities

Assets

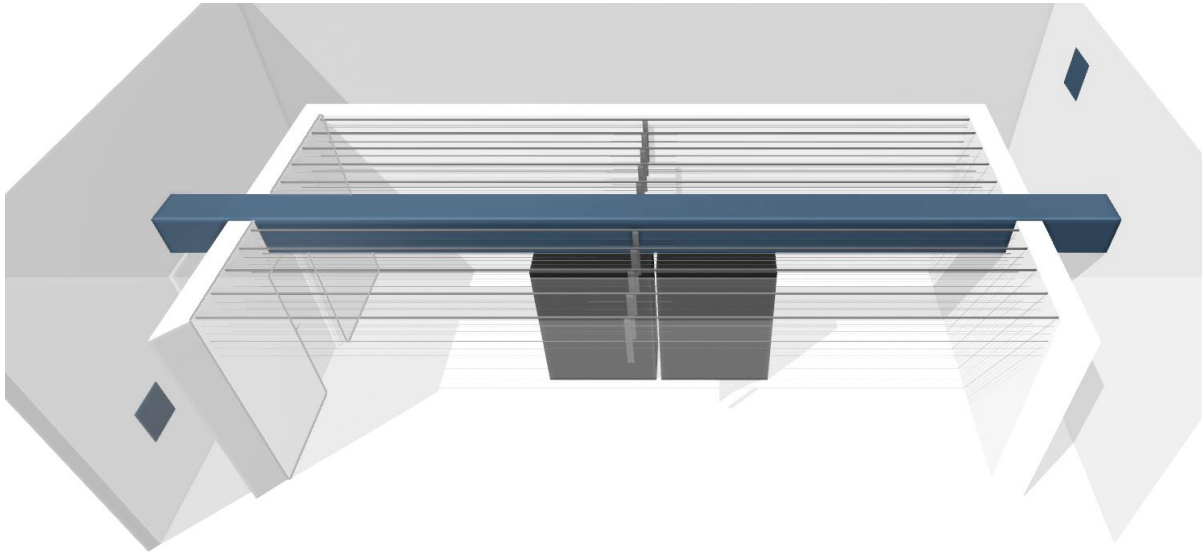
- Existing and future nuclear configurations
- Scale 1
- Modularity of the installation
- Multi-client



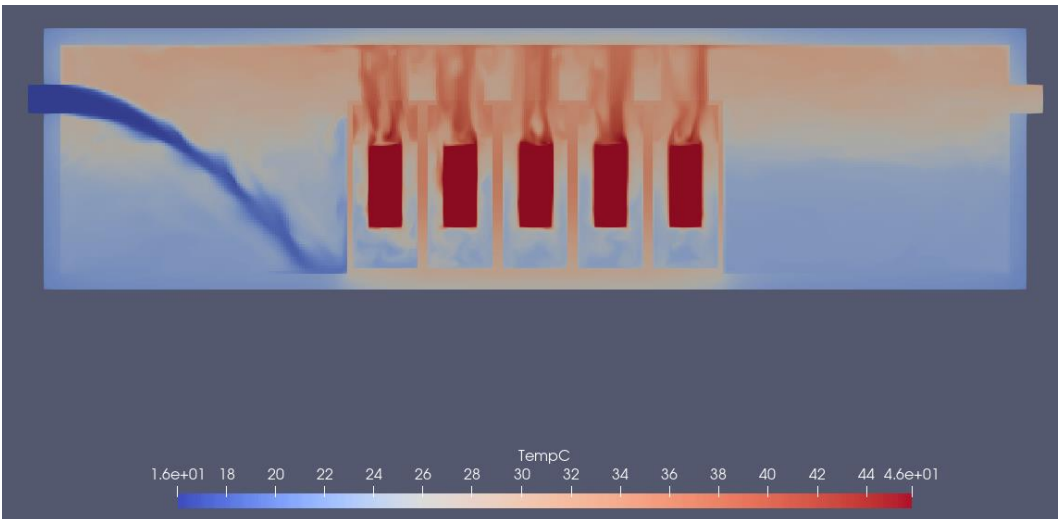
Main technical objectives

- Experimental evaluation of the metrological performance of innovative measurement systems
- Consolidation of numerical calculation codes used in ventilation
- Support for the design and commissioning of ventilation installations
- Optimisation in deconstruction (filtration, air locks)

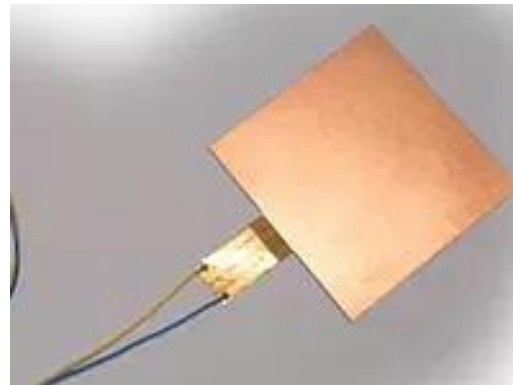
Representative concrete room



- Double enclosure (control of initial conditions in the external skin)
- Representative internal walls (thickness and reinforcement)
- Measurement of all the main thermo-aerodynamic variables:
 - Air flow
 - Flux at the walls
 - Temperature distributed by optical fiber (Raman)
 - Pressure and hygrometry



Rij / user wall function – size : 0,033m



3D numerical experiments – Cases setup

- **Phenomena modelled (3D and 0D)**

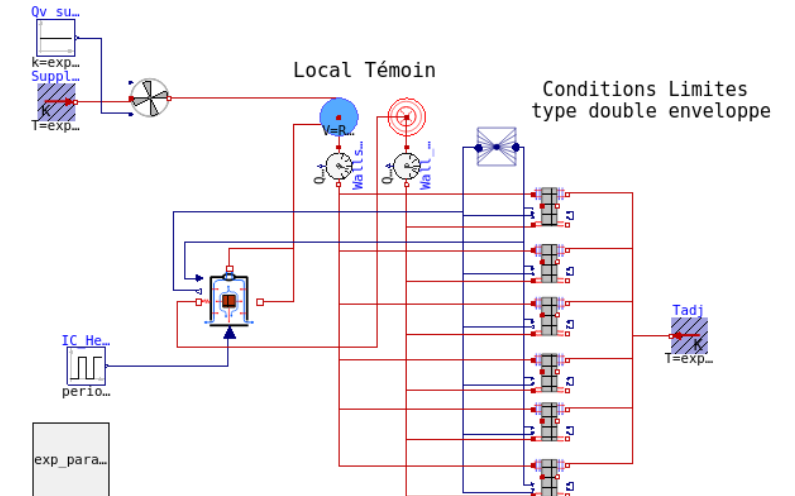
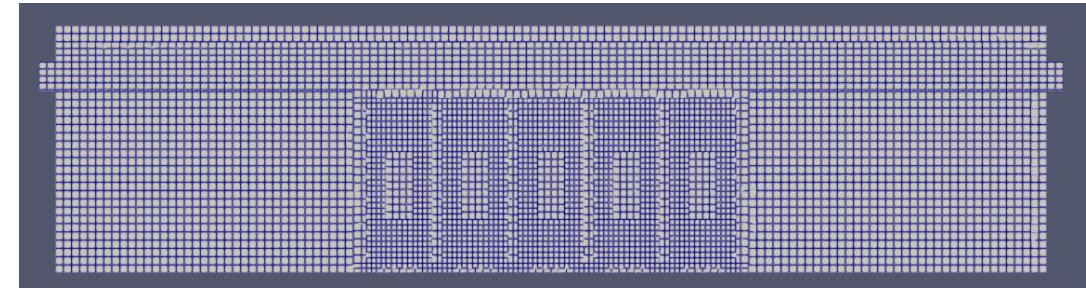
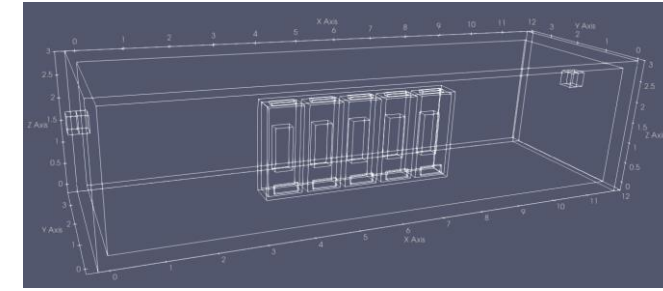
- Fluid flow
- Heat transfers / Conjugate heat transfer
 - Conduction / Convection / Radiant transfers

- **Mesh**

- Air, Walls and Electrical Cabinet Modelled
- Elementary size : 0,1m
- Wall functions : integral correlation used for heat flow

- **Main numerical parameters**

- Unsteady solver
- Turbulence : $k - \omega SST$
- Schemes : upwind for turbulence, centered (with sloped test) for other variables
- Unsteady solver with 0,1 s reference time step
 - Between 3000 and 6000 iterations required
 - Around 10 – 20 h Cpus cumulated time (0,5 – 1h with 20 cores)



3D numerical experiments – Case WorkFlow



- **Cases step**

- Generate mesh
- 3D simulation of air flow and solid conduction

(Salome : geom/smesh)

- 0D/1D Simulation

(Code Saturne)

- Postprocess Results

(TAeZoSyPro with OpenModelica)

- Mean profiles

(Specific Python library)

- Integrate local 3D results

(Specific Code Saturne User functions)

- **Experiences plans**

- Generate Plans

(Persalys)

- Wrap case in _exec function

(Specific Python Library)

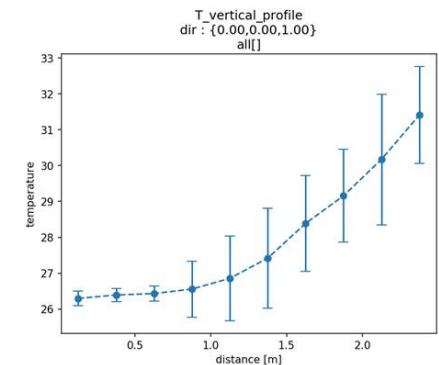
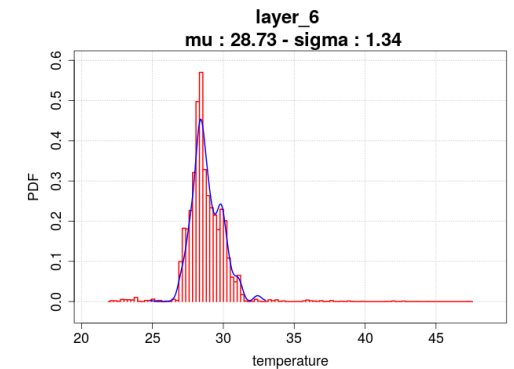
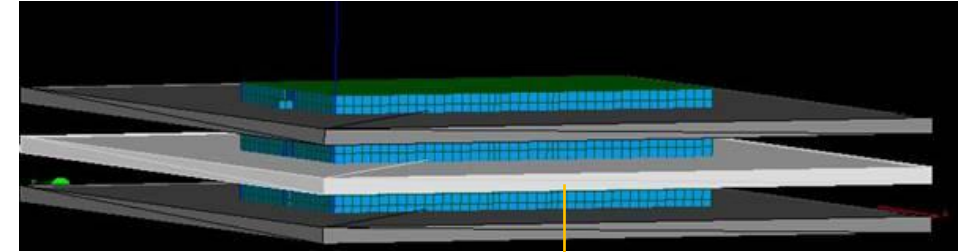
- Enables case restart and results saving

- Evaluate plans until convergency is reached for each case

- Results analysis as a new Data model

- Dependencies

- Sensitivity analyses metamodel based



3D numerical experiments – Persalys WorkFlow

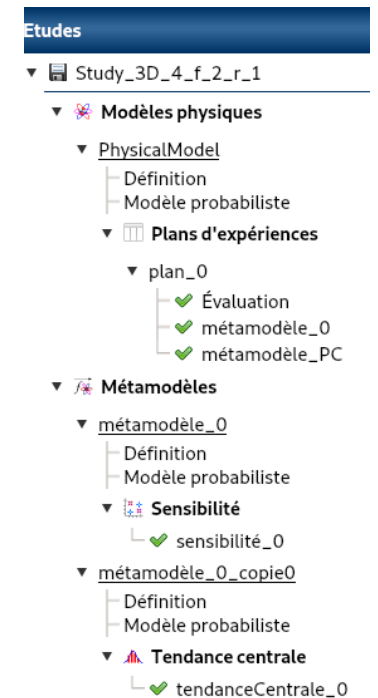
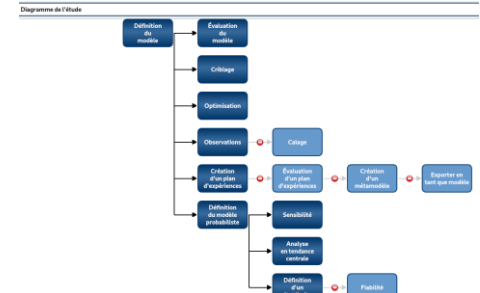


- Persalys _exec function

- Define as a YACS model
- Evaluation of the plan with dumped code modification

```
import salome_ot
import os
study,ModèleYACS_0=salome_ot.getYacsPyStudy(code)
import pydefx
pydefx_path = os.path.dirname(pydefx.__file__)
light_executor_path = os.path.join(pydefx_path, "plugins", "lightexecutor.py")
mybuilder = pydefx.slurmbuilder.SlurmBuilder(executor=light_executor_path)
myModel = pydefx.SlurmStudy(schemaBuilder=mybuilder)
ModèleYACS_0.setJobModel(myModel)
```

- Use 'slurm study' feature to evaluate point by block (10 for instance)
- Use of default YACS model does not allow executables to be launched with *srun* command
- Executable to be launched with : *srun --exclusive -n X ./my_mpi_program*
 - ❑ *--exclusive* options allow slurm to use all allocated resources
 - ❑ *Waste some resources when block size increases : reason not identified*
 - *Example : block size of 10 with 204 cpus allocated on 6 nodes allow evaluations 10 by 10*
 - *Block size of 50 with 1020 cpus on 30 nodes allows evaluations 37 by 37 instead of expected 50*



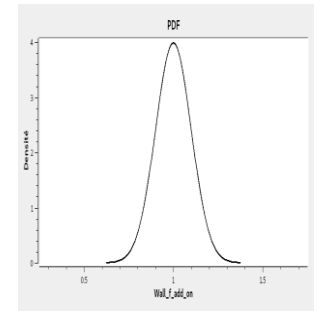
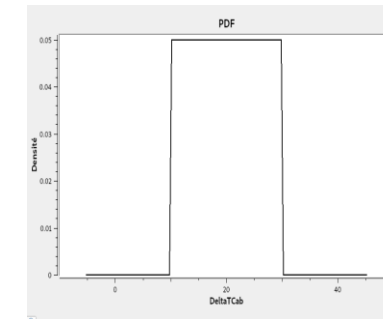
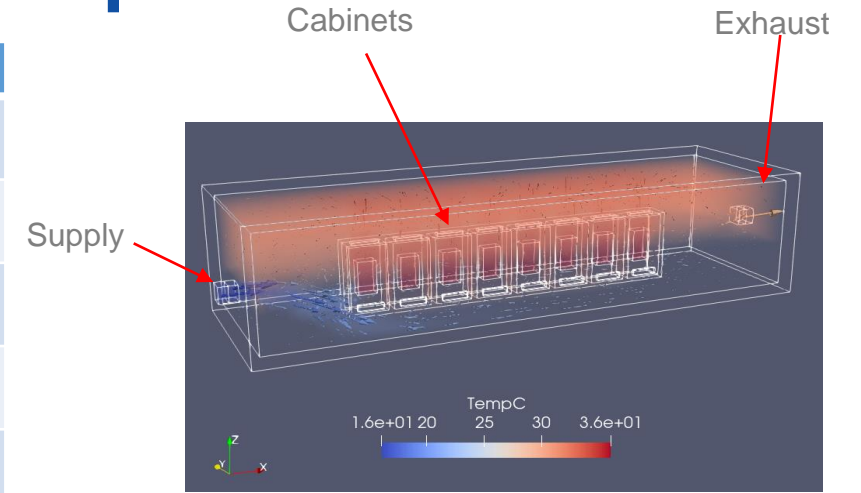
3D numerical experiments – Uncertain parameters

Geometrical parameters

Parameter	Marginal	description
Inlet_Z/Y_percent [-]	Uniform [0,1 – 0,9]	Supply position
Outlet_Z/Y_percent [-]	Uniform [0,1 – 0,9]	Exhaust position
Cab_Em_per [-]	Uniform [0.4,0.7]	Emitter volume percentage of cabinet
Cab_X/Y_Percent [-]	Uniform [0.1,0.9]	Position of cabinet row in the room
HL_W_Cab [W]	Uniform [300,1000]	Heat Loads per cabinet (round to inf int)
Cabinet_H [m]	Uniform [1.5,2]	Cabinet height
Room_H [m]	Uniform [2.4,2.8]	Room height
Renewal_Rate [vol. h ⁻¹]	Uniform [1,10]	Air change rate
Inlet_Speed [m. s ⁻¹]	Uniform [1,5]	Supply air speed
HL_density [W. m ⁻²]	Uniform [50,150]	Heat Loads density
T_supply [degC]	Uniform [16,20]	HVAC supply temperature
T_adj [degC]	Uniform [20,30]	Adjacent room temperature
Wall_f_add_on [-]	Normal [$\mu = 1.0, \sigma = 0.1$]	Modifier of predicted wall heat exchange coefficient with integral correlation
KsiCab [-]	Uniform [6-50]	Number of dynamic pressure losses (at inlet speed) of Cabinet

Limit conditions

Model param



- **Methodology**
 - Study equiprobable situations
 - Investigate generic model sensitivity
- **Dependency**
 - Independent Copula

3D numerical experiments – Variables of interest

Power balance

- Radiant/convective heat flow for surface
- HVAC power
 - $P_{HVAC} = \dot{m}c_p(T_{supply} - T_{Room})$
- Convergency at final time step
 - $cvg = \frac{P_{instationnaire}}{Heatloads}$

Temperatures

- $T_{moy} \text{ 3D} / T_{0D} / T_{outlet \text{ 3D}}$
- $T_{inlet} / T_{outlet} \text{ cabinets}$
- Vertical mean profile temperature:
 - 10 layers : $\mu_{layer_id}^* = \frac{\sum_{c_id=0}^{n_cells-1} w_{c_id} \times x_{c_id}}{\sum_{c_id=0}^{n_cells-1} w_{c_id}}$
 - With : w_{c_id} weight of cell for layer id

Scales

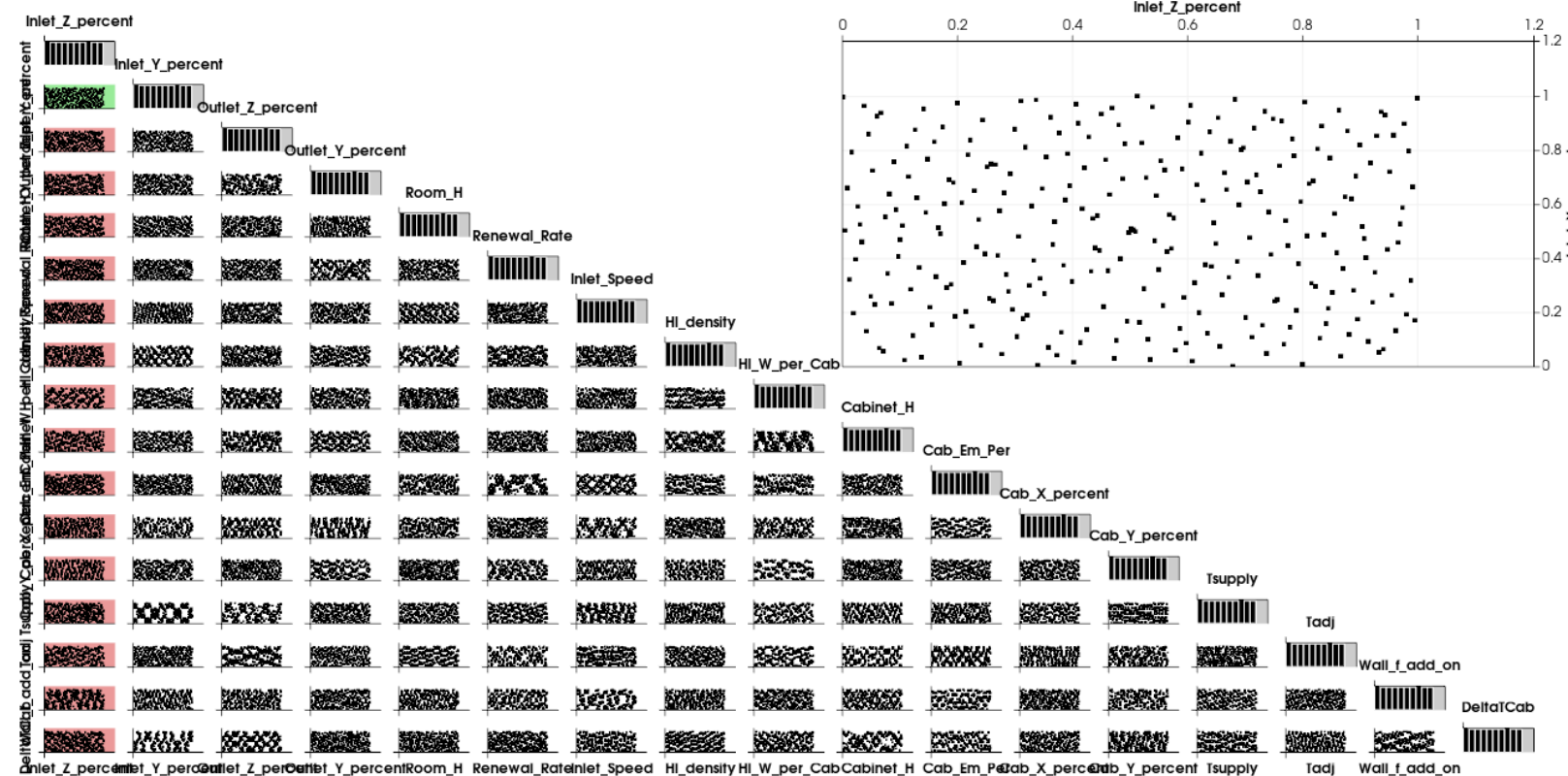
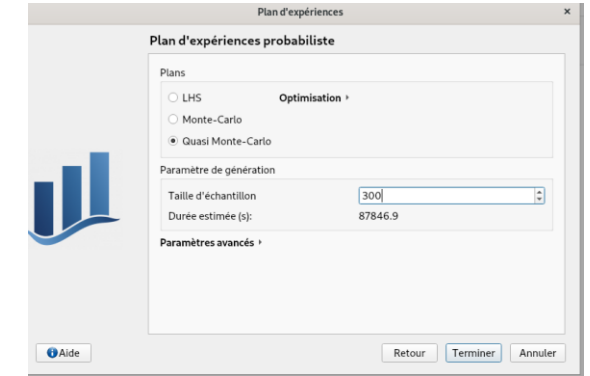
- Power : $P_{heatloads}$
- Dimensionless power : $\frac{P}{P_{heatload}}$
- Temperature $\begin{cases} \frac{1}{2}\xi\rho_0V^2 = \Delta\rho gh = \rho_0\beta\Delta T_{ref}gh \\ \dot{m}c_p\Delta T_{ref} = P_{th} \end{cases} \Rightarrow \begin{cases} V = \left[\frac{2\beta P_{th}gh}{S_{cab}c_p\xi\rho} \right]^{\frac{1}{3}} \\ \Delta T_{ref} = \frac{P_{th}}{\rho V S_{ar}c_p} \end{cases}$
 - with $\xi = 1 / S_{cab} = 0.07$: cabinet inlet surface
- Dimensionless temperature : $T_n = \frac{T - T_{mean \text{ 3D}}}{\Delta T_{ref}} / \Delta T_n = \frac{\Delta T}{\Delta T_{ref}}$

Methodology

- **Numerical integration** of local 3D variables
- **Project 3D field on a fixed 1D field** (mean profile)
- Get **similar** variables of interest between **0D** and **3D** codes
- **Dimensionless** variables to help **comparison** between cases

3D numerical experiments – generated experience plan

- Probabilistic experience plan : Low discrepancy experience plans (Quasi Monte Carlo)
 - 300 points
 - 17 uncertain parameters
 - Sensible filling rank space given the use of uniform and independent marginal law
 - Try to fill optimize space filling (distance between two points)
- Point number : Compromise between computing time and number of points



3D numerical experiments – Results

TempC distribution for each profile layer

Persalys run report

62 Study 1/OT Persalys 00105

Point Values

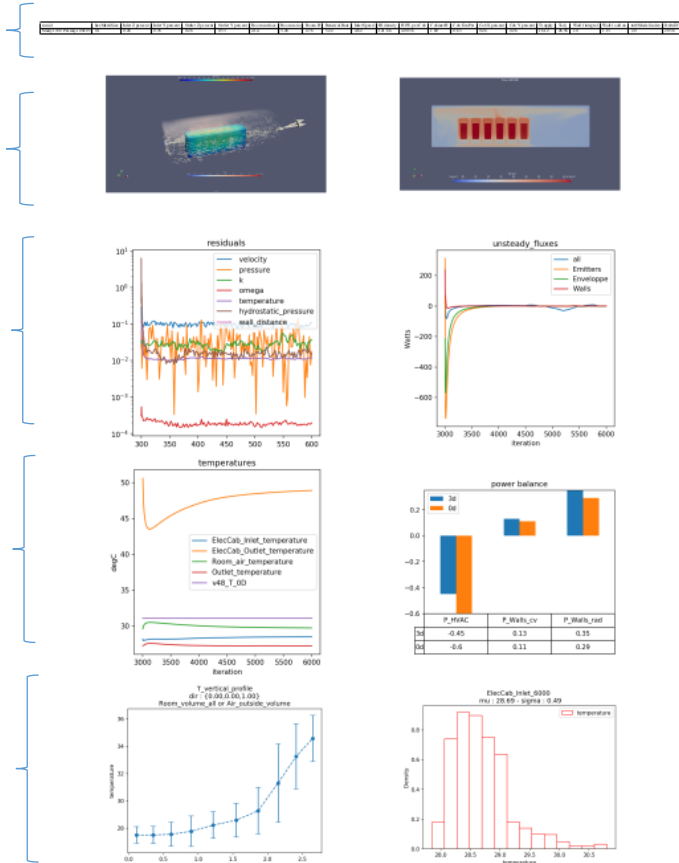
Catalyst View

Convergence

HVAC balances
Temperature – Power
Balance
(0d/3d)

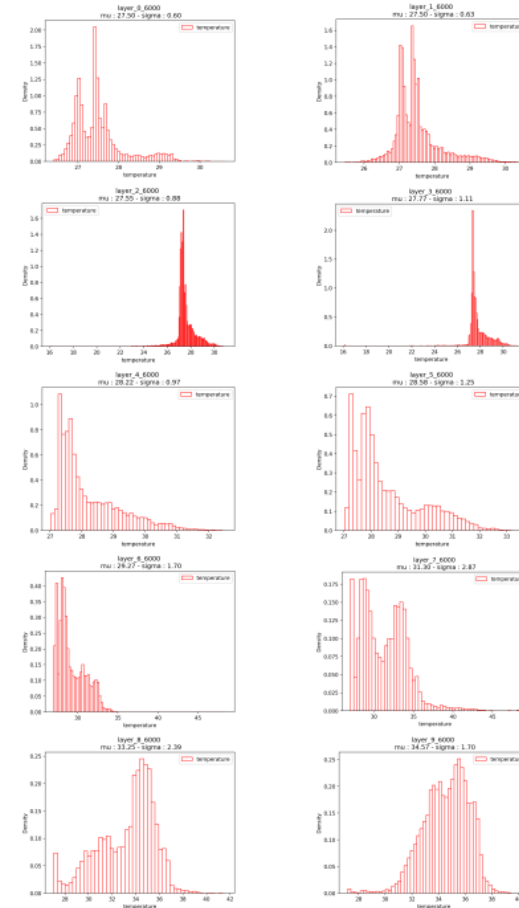
Mean TempC profile

TempC distribution at
cabinet Inlet



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Persalys run report



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- **Total evaluations**
 - First experience plan of 16 cases with different numerical setups : 10 000 Cpus days
 - Determine sensible compromises between precision and cpu cost of each point evaluation
- **Evaluations**
 - 18 points failed due to 3D convergence issues
 - Cpus Time : 140 cumulated days – 1.5 day (user) on GAIA
- **Report**
 - LaTeX PDF report
 - Run_id exported as variable of interest to make the link between Persalys experience plan table and detailed run results

3D numerical experiments – Results

- **Reminder : implication of 0D approach**

$$\sum_i^{supply} (\dot{m}_i c_{p_i} T_i) - \dot{m}_{exhaust} c_{p_{room}} T_{air} + \sum_i^{walls} h_{total} S_{wall} (T_{wall} - T_{air}) + W = 0$$

- Local energy equation integrated on the whole room volume

$$\frac{\partial \hat{T}}{\partial \hat{t}} + \hat{\rho} \hat{V} \cdot \hat{\nabla} \hat{T} = \frac{1}{\sqrt{RaPr}} \hat{\nabla} \hat{\lambda} \hat{\nabla} \hat{T}$$

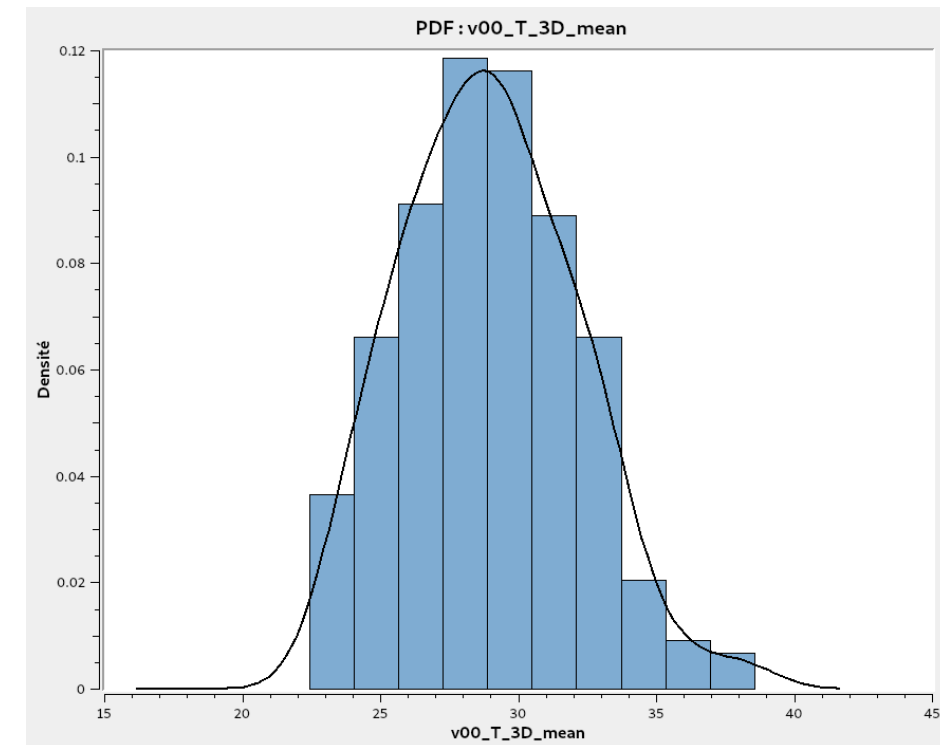
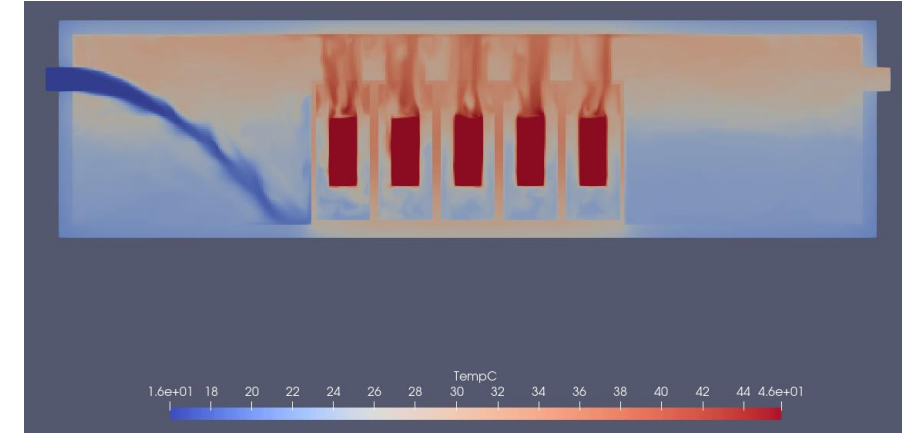
- No reason, given expected heterogeneity that T_{air} is the relevant potential in the first equation, $T_{air} = \frac{1}{m_{air}} \iiint_V \rho T dV$

- **Analysis focused :**

- Stratification phenomena (given by the 3D code)
- Consistency between predictions of 3D and 0D code

- **Global results**

- Sensible global mean room temperature distribution given the range of the uncertain parameters



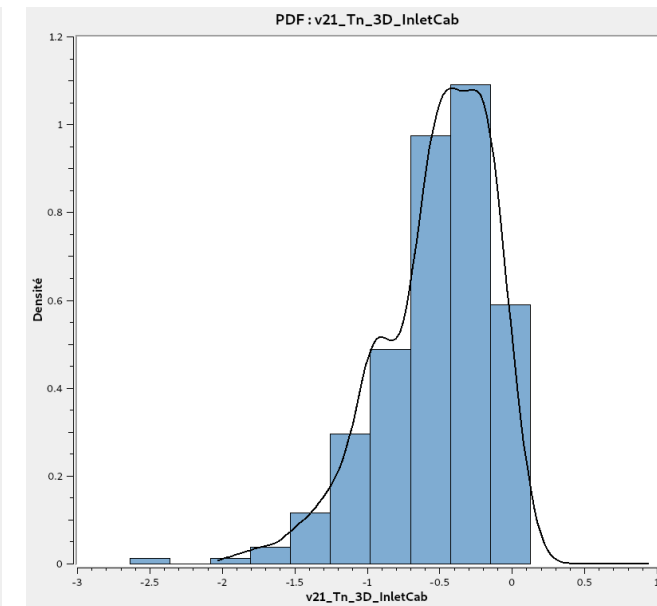
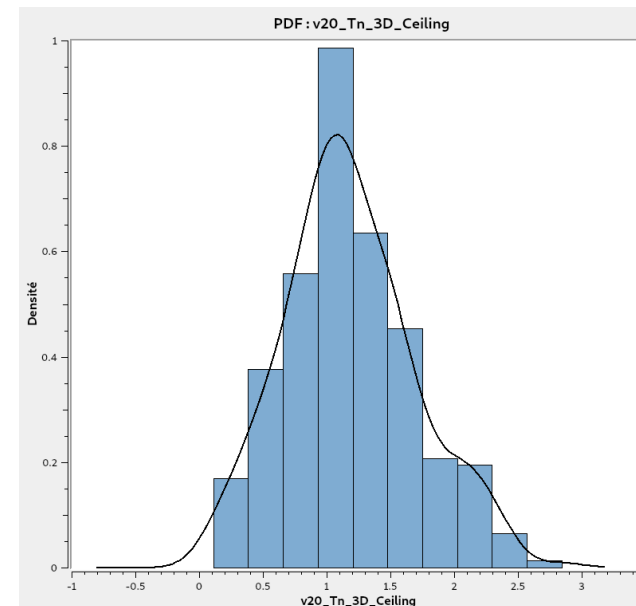
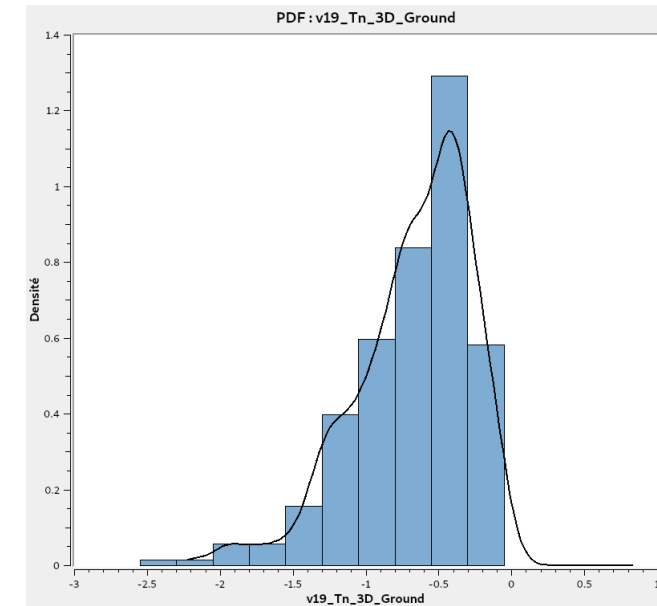
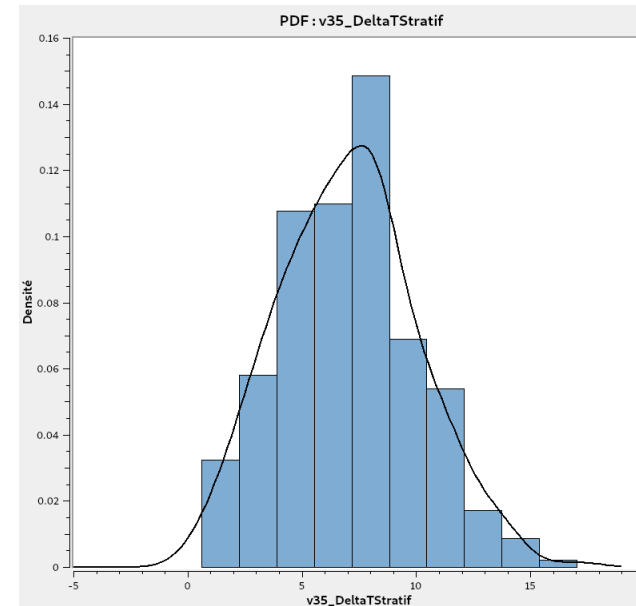
3D numerical experiments – Results

- Main dependencies (Spearman indices)

- $\Delta T_{stratification}$: -0,48 Inlet_Z_percent / -0.469 : Inlet_speed / 0,428 : HL_density / 0,179 HL_W_per_Cab / 0,2 Tadj /
- Tn_{sol} :0.483 Inlet_Z_percent / 0.214 Renewal_Rate / 0.42 Inlet Speed / -0.434 HL_density / 0.334 HL_W_Per_Cab

- Analysis

- Stratification tends to be mainly linked to supply parameter (geometry and way of supply) combined with the need of a certain amount of heat loads
- Emitters parameters are less influent once heat loads density fixed
- Emitter inlet temperature always below mean air temperature
 - Cabinets act as an active devices with air inlet on the bottom part and blowing air in upper part of the room

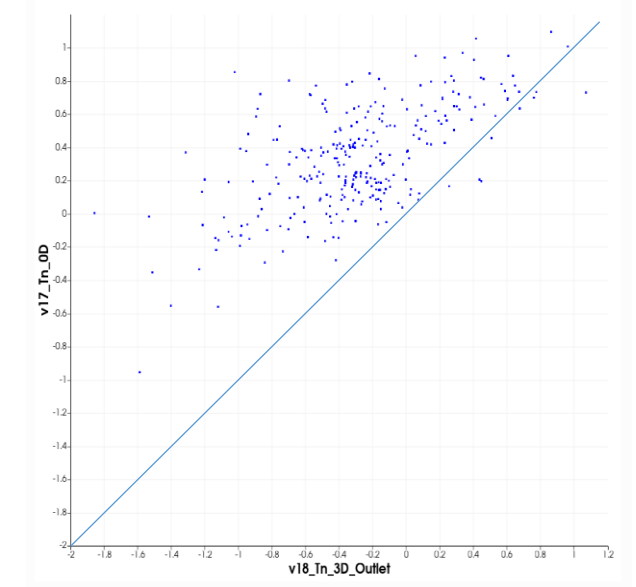
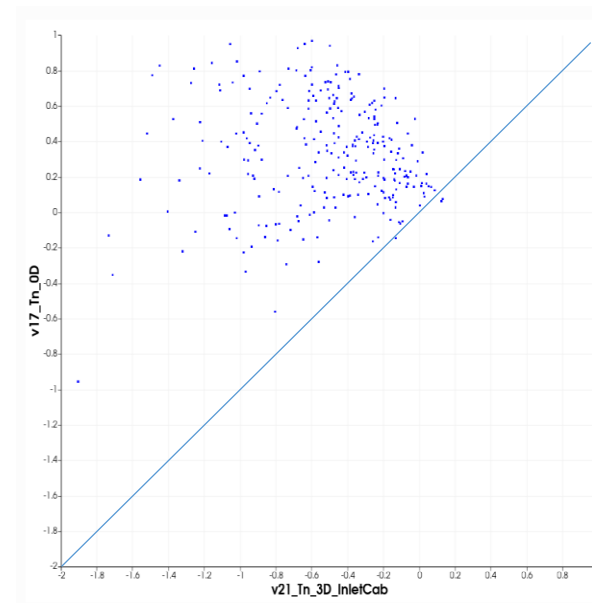
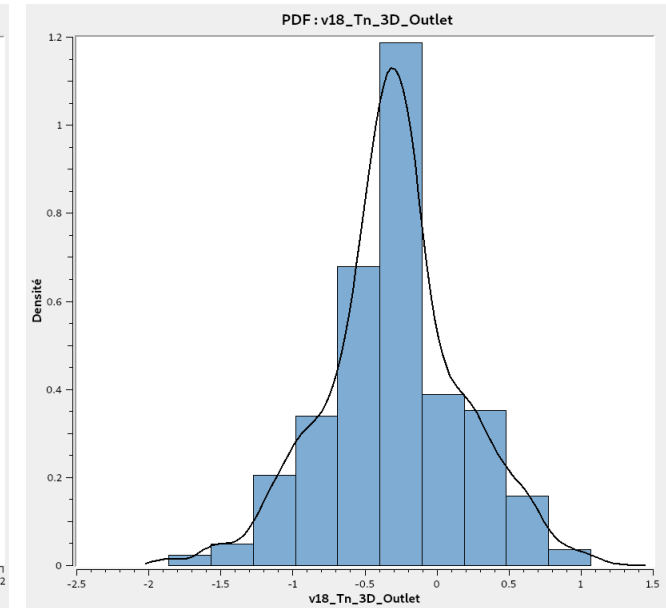
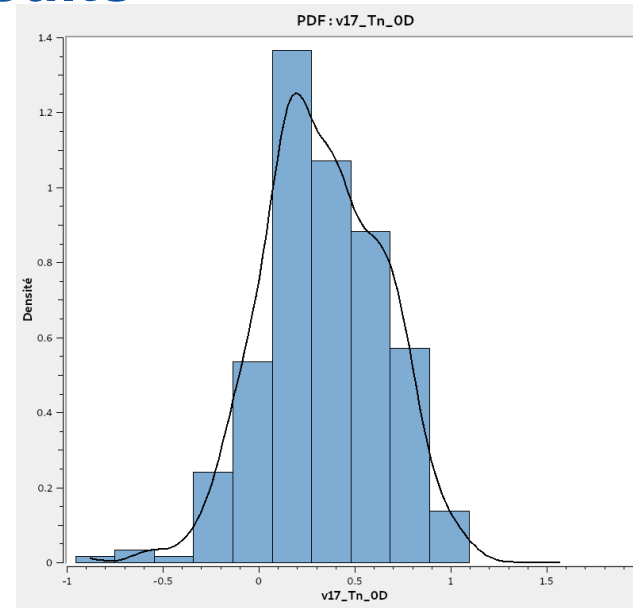


3D numerical experiments – Results

Temperatures : 0D – 3D

- Analysis

- T_{0D} globally above $T_{CabInlet}$
 - Case or both are close are linked to case with low absolute temperature stratification
- Over estimation of HVAC power with 0D approach (link to poor ceiling heat transfer estimation)
- Cases with T_{0D} lower than $T_{3D Mean}$ combined stratification and exhaust in lower part of the room



3D numerical experiments – Results

MetaModels

- **Objectives**

- Learn dimensionless temperature to
 - Perform quantified sensitivity analysis (Sobol')
 - Perform partial quantified sensitivity analysis (with some influent parameters set as constant)
 - Allow metamodel export for improve 0D modelling of electrical premises

- **Technics used**

- Polynomial chaos
- Kriging

- **Validation criteria**

- $Q^2 = 1 - \frac{\sum_{j=1}^{n_t} (Y_j - \hat{Y}_j)^2}{\sum_{j=1}^{n_t} (Y_j - \bar{Y})^2}$ / Compute with test sample, analytic method and K-Fold
- Qualitative Analysis of residuals distribution
 - Aim is to have a metamodel averagely accurate

3D numerical experiments – MetaModels - Kriging

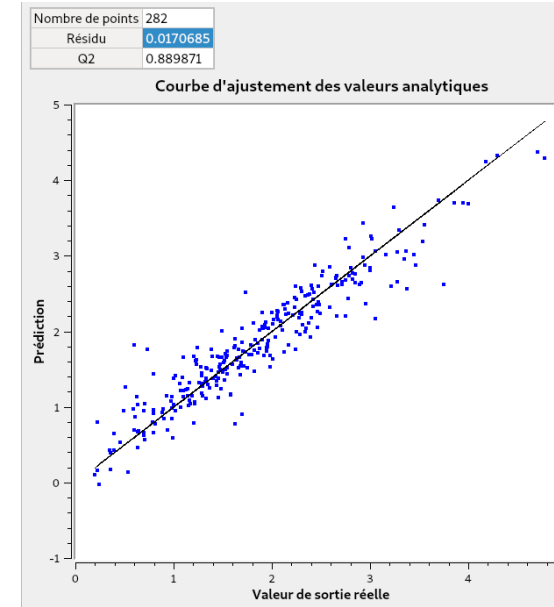
- Parameters

- Covariance kernel : squared exponential
- Constant tendency with optimize hyper parameters ΔT_n

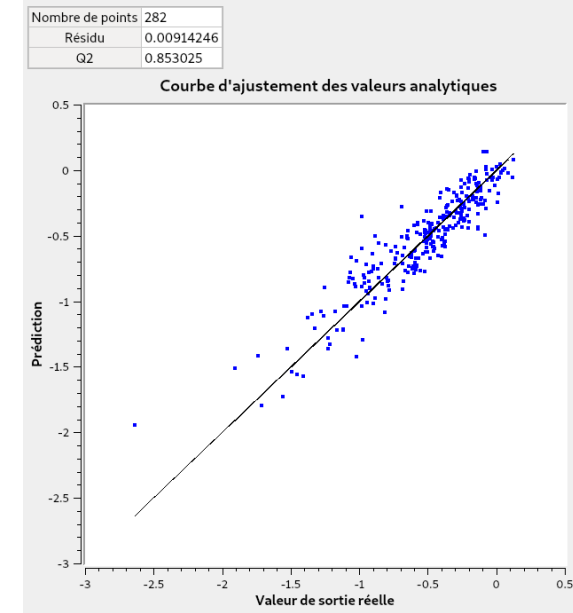
- Results

- All scales within bounds except Wall_f_add_onn (less influent parameter)
 - Scales for normalized parameter $\frac{x - \bar{x}}{sd(x)}$
- Acceptable Q2 for 3D based variables

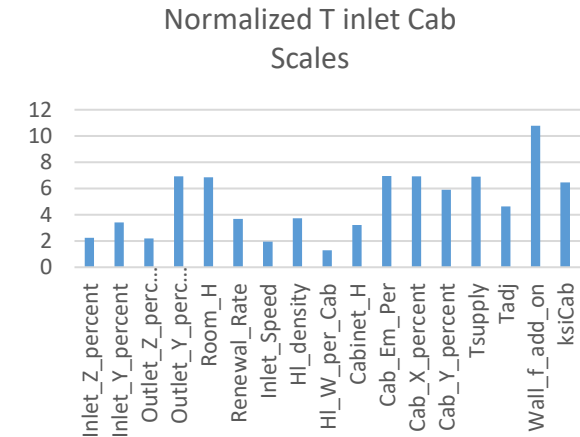
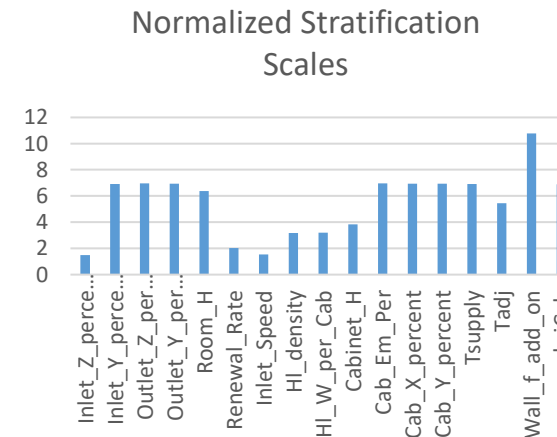
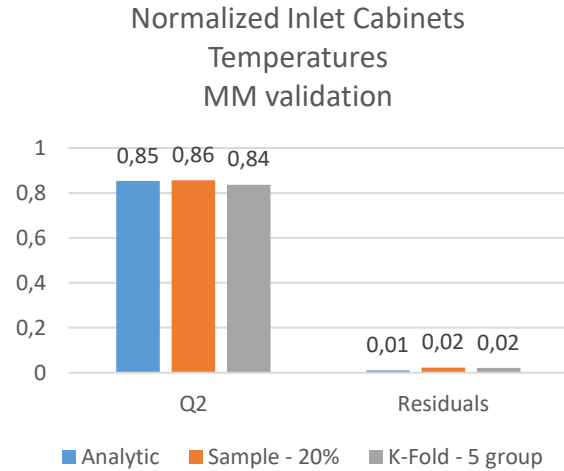
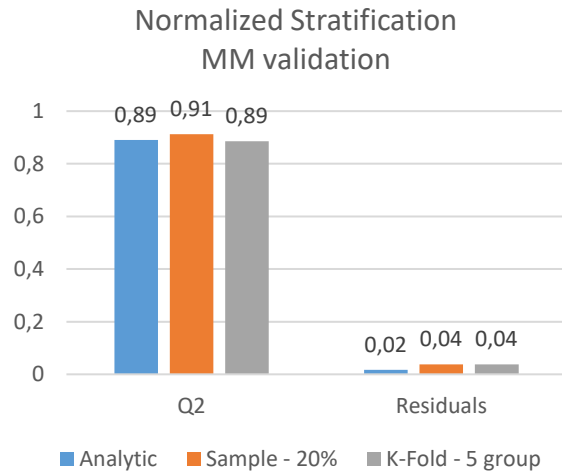
- Validation



ΔT_n – Stratification



T_n – Inlet Cabinet



3D numerical experiments – MetaModels – Kriging

Sensitivity

- **Method**

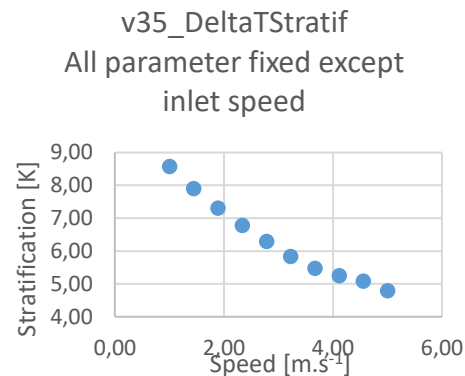
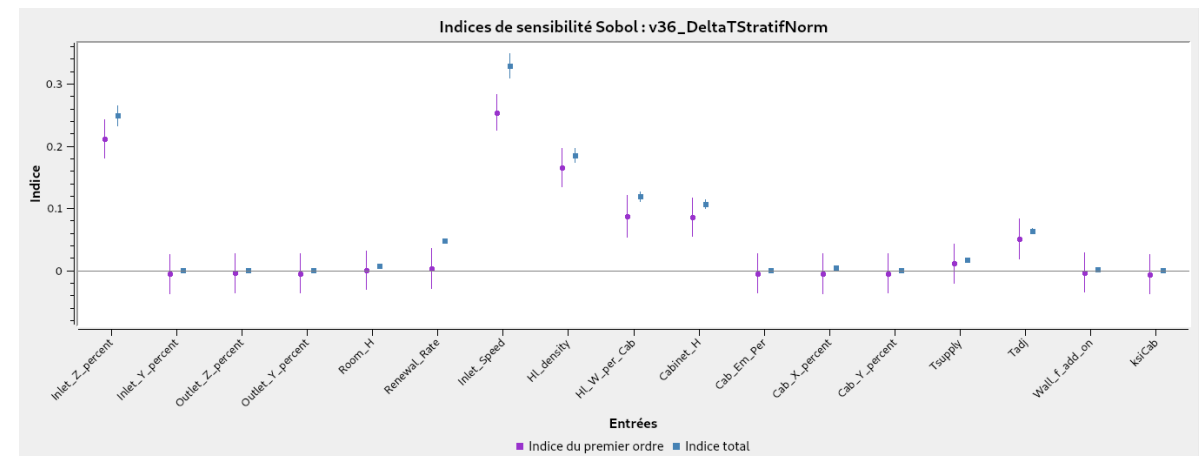
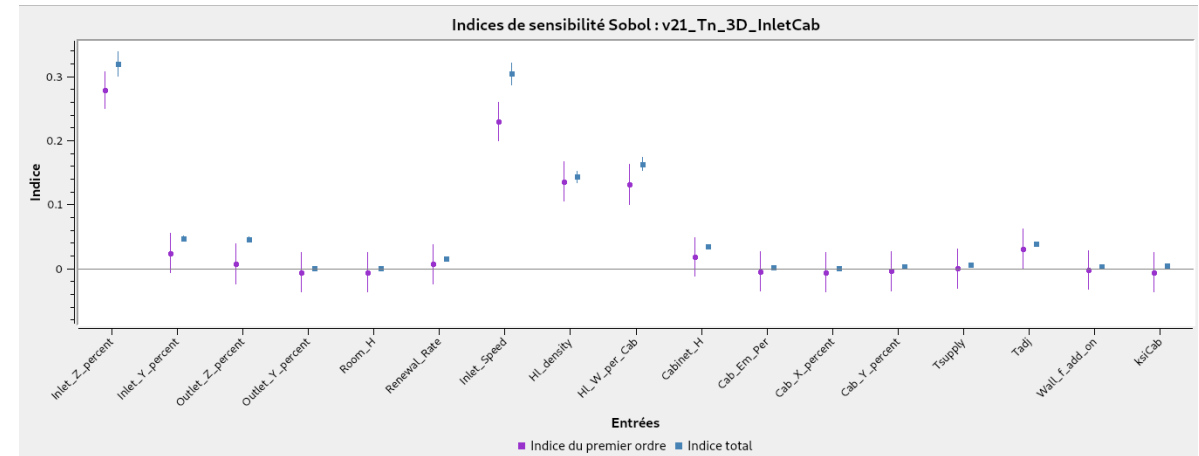
- Use generated MetaModel to estimate Sobol' indices

- **Analysis**

- Results consistent with Spearman indices and computed kriging scales
- Significant role of interactions
 - Supply geometry and inlet speed and air change rate
 - Deeper investigation of supply geometry and parameters to be done on next experience plans (supply angle, jet turbulence, inlet grid roles ...)
- Can be used to generate experience plan of lower dimension
 - Projected 1D (all parameters fixed except 1) often tends to have a kriging variance nearly equal to amplitude parameter
 - Number of points for model training is low given the input dimension

$$S_i = \frac{Var[E[Y|X_i]]}{Var[Y]}$$

$$ST_i = 1 - \frac{Var[E[Y|X_{\sim i}]]}{Var[Y]}$$



3D numerical experiments – MetaModels – Polynomial Chaos

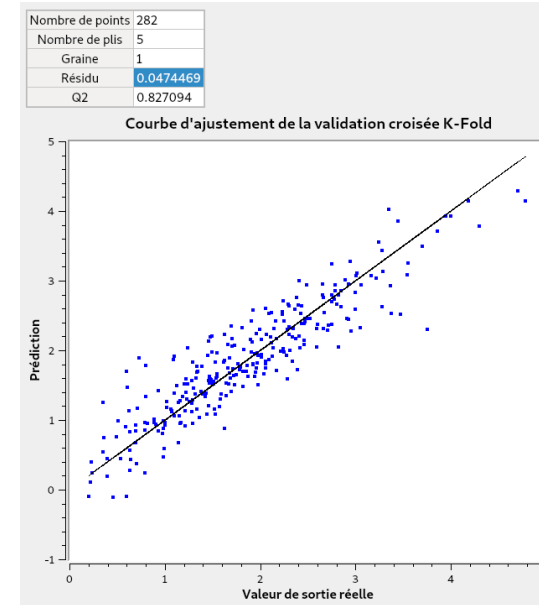
- Parameters

- Max degree : 2 / Sparse basis

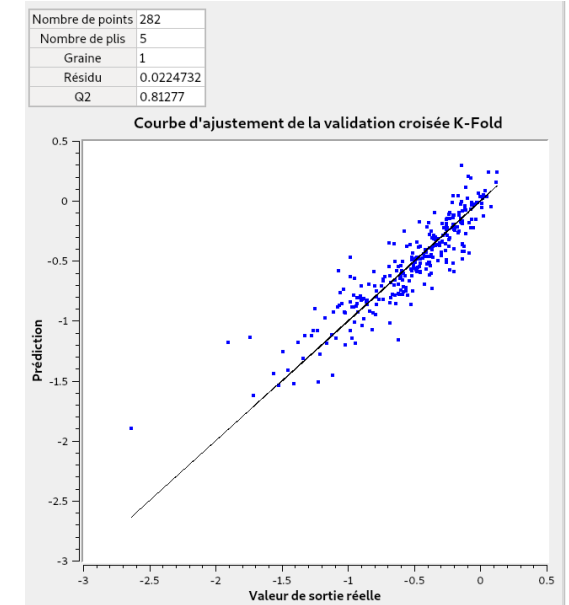
- Results

- Slightly lower Q2 than kriging method
 - Lower maximum degree based on explained part of variance
 - Specify marginal of input samples improve significantly Q2

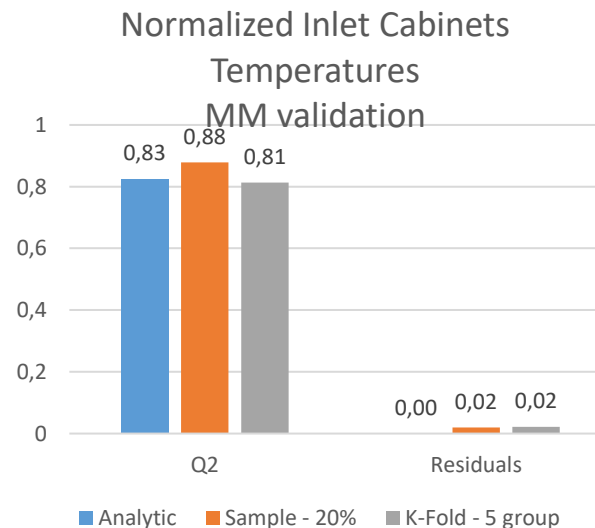
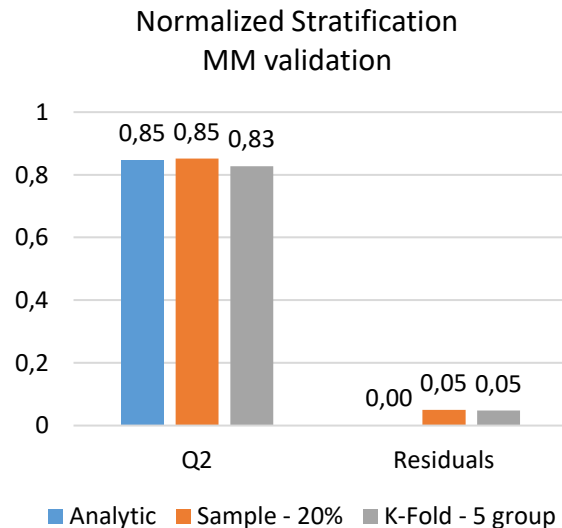
- Validation



ΔT_n – Stratification



T_n – Inlet Cabinet



3D numerical experiments – MetaModels – Kriging

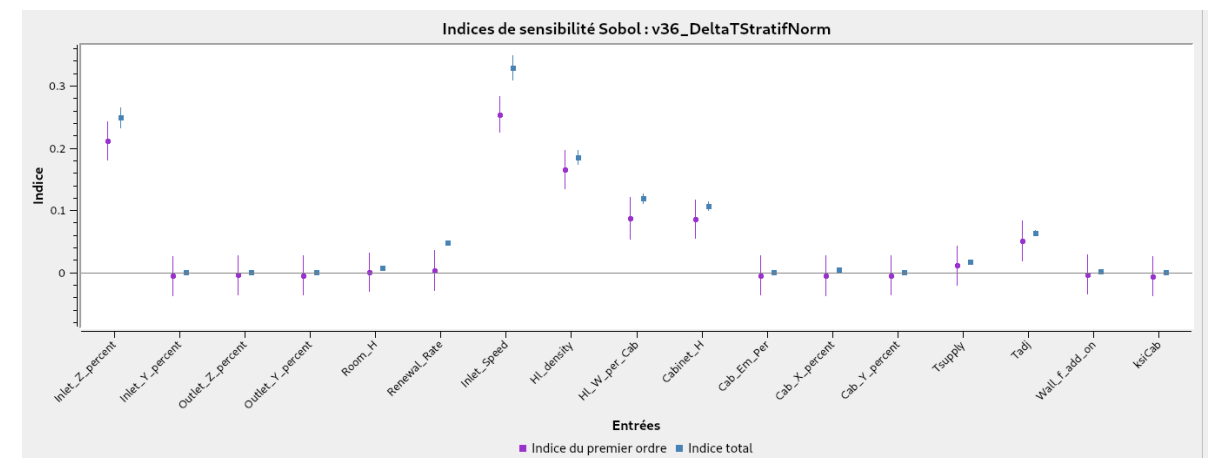
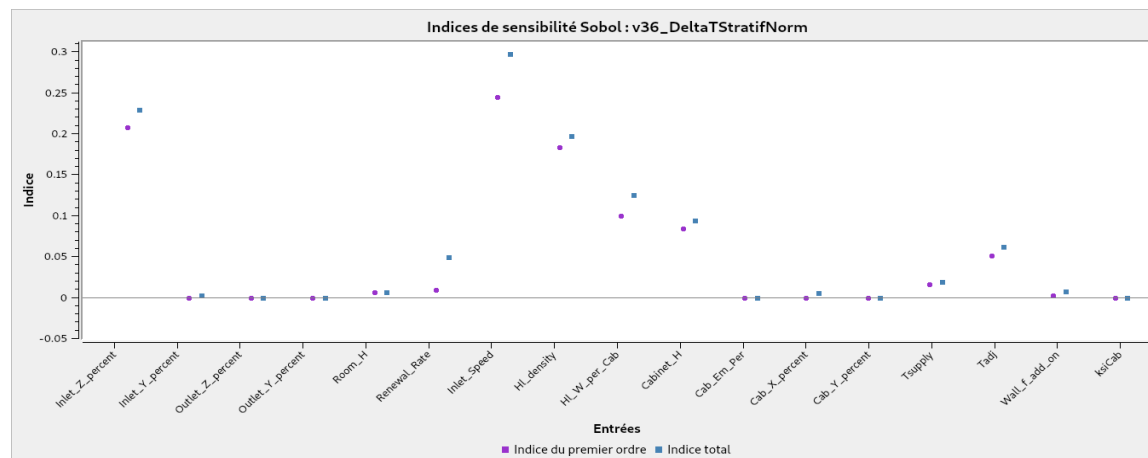
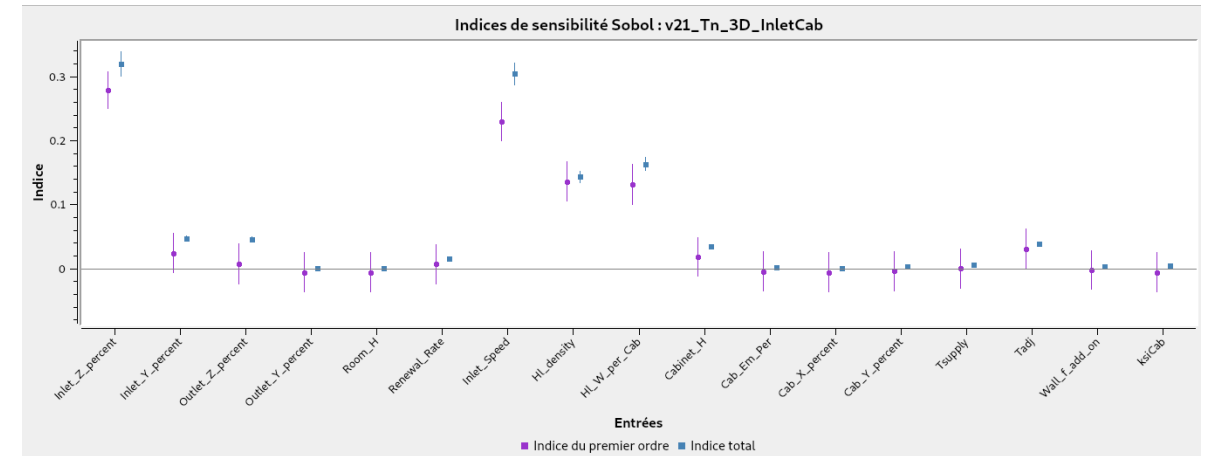
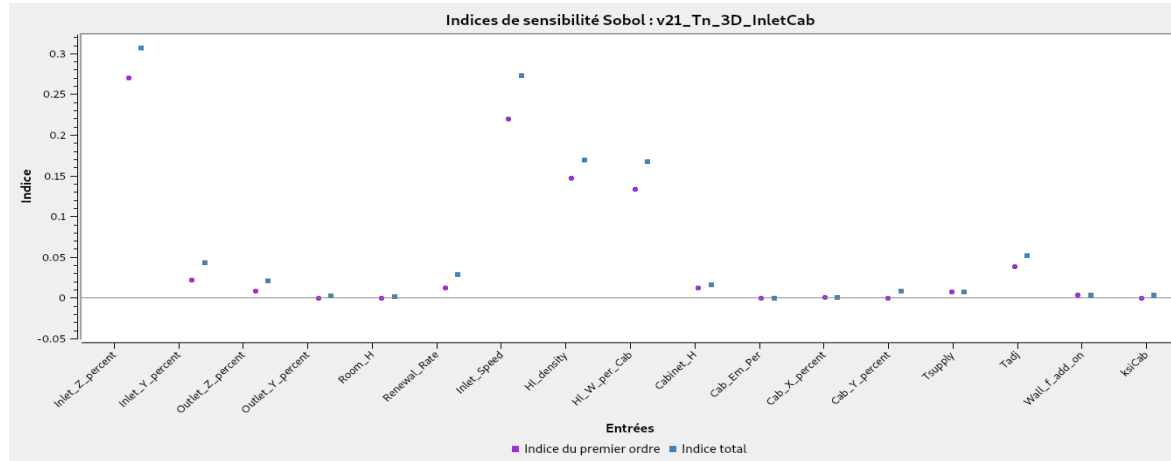
Sensitivity

- Analysis

- Consistent Sobol indices between the two models

$$S_i = \frac{\text{Var}[E[Y|X_i]]}{\text{Var}[Y]}$$

$$ST_i = 1 - \frac{\text{Var}[E[Y|X_{\sim i}]]}{\text{Var}[Y]}$$



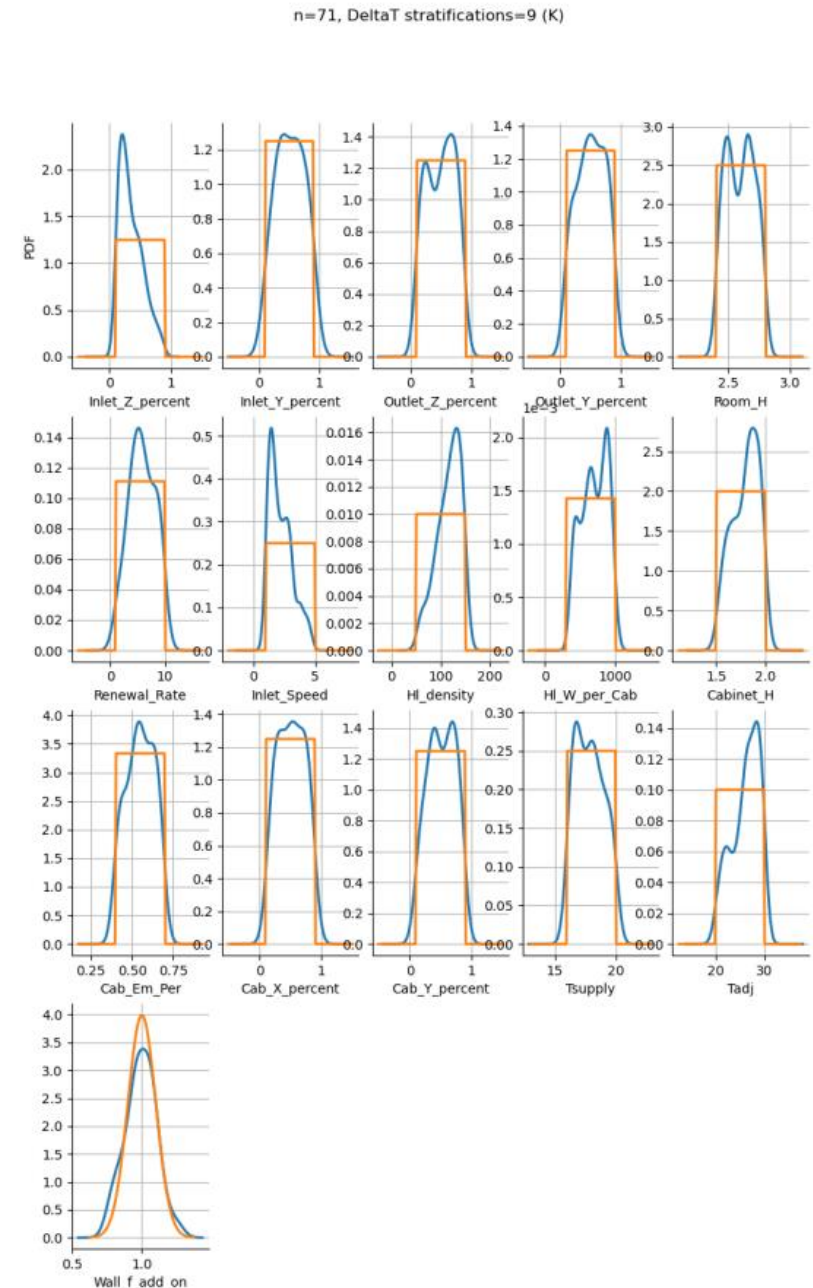
3D numerical experiments – Estimate conditional marginal law¹

- **Method**

- Select a sample from the evaluated experience plan based on quantile of a variable of interest
- Let Y_s be a threshold on the output, e.g. computed from a quantile of Y . Plot the unconditional distribution of each marginal input X_i and the conditional distribution $X_i | Y > Y_s$, where If the variable X_i is independent from the event $X_i | Y > Y_s$, then the two distributions are equal. Otherwise, there is a dependency
- Chosen variable of interest : $\Delta T_{stratification}$

- **Analysis**

- A significant stratification modify the input parameter distribution for the most influent parameters
- These parameters are the same than the ones previously identified (with Spearman indices and then Sobol' indices estimation with MetaModel)



Conclusions and perspectives

- **Conclusion**

- **Tool chain operational** for thermo aerolic sensitivity studies : Based on Salome platform
- **First quantification** of influent parameters leading to air mix in industrial premises
 - **Significant** role of **supply** : need further investigation on its modelling
- **MetaModelling** of integrated local 3D variable with reasonable accuracy
 - CFD with $y^+ = 1$: 5000 cpus days -> 3D with integral correlation : 20 h cpus -> MetaModel : 0,001 s
- **0D** model generally **overestimate** temperature of interest due to the perfect air mix hypothesis



- **Perspectives**

- Select **group of influent** parameters (optimal size base on first or total Sobol' indices) and re-evaluate experience plan with a lower input dimension
- **Implement MetaModel** of stratification in **0D** tool by export it and re import in Modelica Model (work in progress with Phimeca – python.h based for wrapping python object in C function)
 - Quantify error propagation in 0D model based on MetaModel residuals distribution
- Perform dimensional analysis (Vaschy-Buckingham theorem) and try to calibrate traditional π functions



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- Michael Baudin / Ovidiu Mirescu : For their advice in the use of Persalys/OpenURNS and their uncertainties expertise



Thank you for your attention



ANNEXE

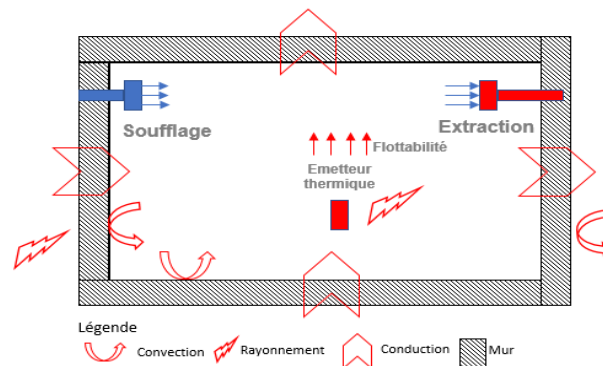
Rappels de la méthode TIPPI (PIRT)

- **L1.01 d:**

- Proposer une identification à priori des phénomènes influents sur la base d'une approche théorique
- Réaliser une première quantification sur la base d'expériences numériques s'inspirant de la géométrie du local Zephyr
- Approche itérative

- **Périmètre**

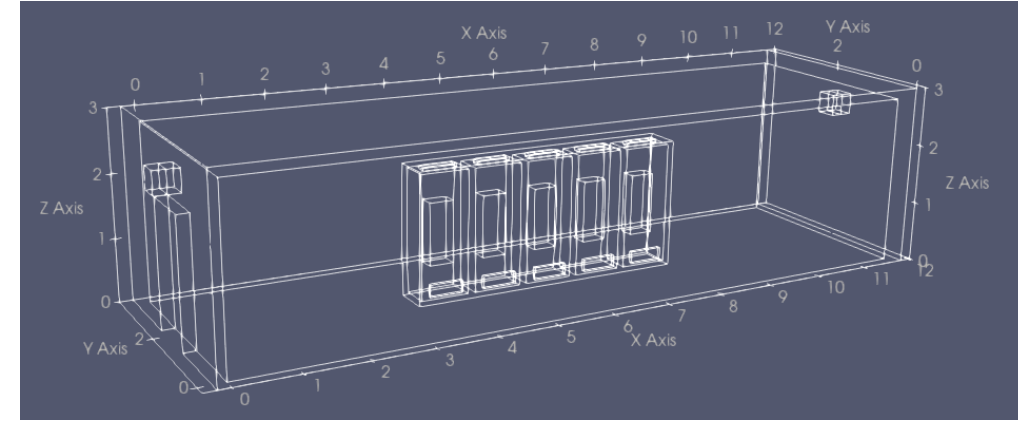
- Problématique vaste et multiphysique
- Les situations thermo aérauliques de locaux industriels nécessiteraient plusieurs PIRT
- Focus on steady state with Zephyr test room laboratory configuration



- 1 Définition de l'étude et objectif du PIRT
- 2 Définition du transitoire et du domaine de simulation
- 3 Identifier les paramètres d'intérêt (Figures of Merit)
- 4 Identifier les principaux phénomènes physiques (schéma simple)
- 5 Lister les paramètres associés aux phénomènes physiques en jeu
- 6 Classer les paramètres par importance d'influence (sur FoM) et selon le niveau de connaissance : Formalisation d'un tableau PIRT
- 7 Procédé itératif : reclasser après de nouveaux résultats

Order of magnitude of the studied case

- Pour le cas de locaux industriels, les ordre de grandeurs sont:
 - $V_0 = 1 \text{ m.s}^{-1} / L = 1 \text{ m} / \rho_0 = 1 \text{ kg.m}^{-3} / \delta p_0 = 1 \text{ Pa}$
 - $\Delta T = 10 \text{ degC} / \lambda_{\text{metal}} = 50 \text{ W.m}^{-1}.\text{K}^{-1} / \lambda_{\text{beton}} = 2.3 \text{ W.m}^{-1}.\text{K}^{-1}$
 - $h_{cv} = 3 \text{ W.m}^{-1}.\text{K}^{-1} / e_{\text{metal}} = 0,05 \text{ m} / e_{\text{beton}} = 0,3 \text{ m}$
- Ce qui conduit
 - Pour les parois : $Ra \approx 1e9 - 1e10$
 - $Ri \approx 1$ loin de singularité
 - $Re \approx 1e5$: écoulement turbulent, notamment au niveau des jets
 - $Pr \approx 0,7$
 - $Bi_{\text{metal}} \approx 3e^{-3} / Bi_{\text{beton}} \approx 3e^{-1}$
- Ces nombres seront à réévaluer de manière plus précises selon les cas étudiés ; les valeurs mentionnées sont utiles pour situer le problème



$$\nabla \cdot \rho \underline{V} = 0$$

$$\frac{\partial \hat{\rho} \hat{\underline{V}}}{\partial \hat{t}} + \hat{\underline{V}} (\hat{\rho} \hat{\underline{V}}) = -Eu \hat{\underline{V}} \hat{p} + \frac{1}{Re} \hat{\underline{V}} \hat{\underline{\sigma}}' - Ri \underline{e}_z$$

$$\frac{\partial \hat{T}}{\partial \hat{t}} + \hat{\rho} \hat{\underline{V}} \cdot \hat{\underline{V}} \hat{T} = \frac{1}{\sqrt{RaPr}} \hat{\underline{V}} \hat{\lambda} \hat{\underline{V}} \hat{T}$$

$$\hat{p} = \frac{p - \rho_0 g \cdot x}{\delta p_0} / \hat{\rho} = \frac{\rho}{\rho_0} / \hat{t} = \frac{V_0}{L} t / \hat{x} = \frac{x}{L} / \hat{\underline{V}} = \frac{V}{V_0} / \hat{\underline{V}} = L \underline{V}$$

$$Eu = \frac{\delta p}{\rho_0 V_0^2} / Re = \frac{\rho_0 V_0 L}{\mu_0} / Ri = \frac{g \beta \Delta T L}{V_0^2}$$

$$: Pr = \frac{\mu_0 c_p}{\lambda_0} / Ra = Gr Pr = \frac{\rho c_p g \beta \Delta T L^3}{\lambda v} / Gr = \frac{\rho g \beta \Delta T L^3}{\mu} \propto Re^2$$

3D numerical experiments – tools used



- Numerical tools used

- Code Saturne (CFD)
- Salome platform
 - Geom/Smesh Modules(CAO/meshing)
 - Persalys Module (YACS model)
- OpenModelica with TAeZoSysPro library
 - Use of ModelicaScripting library to wrap OpenModelica Model in Persalys _exec function

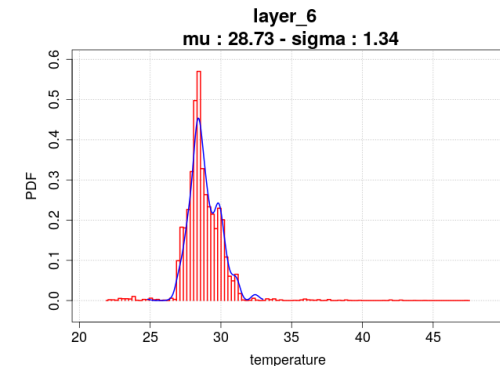
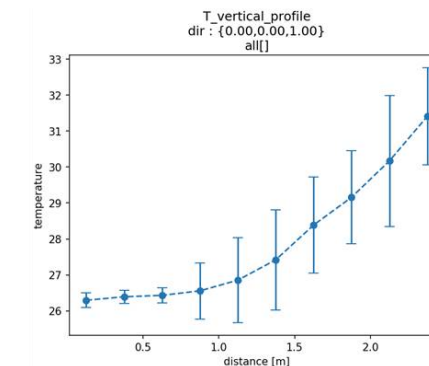
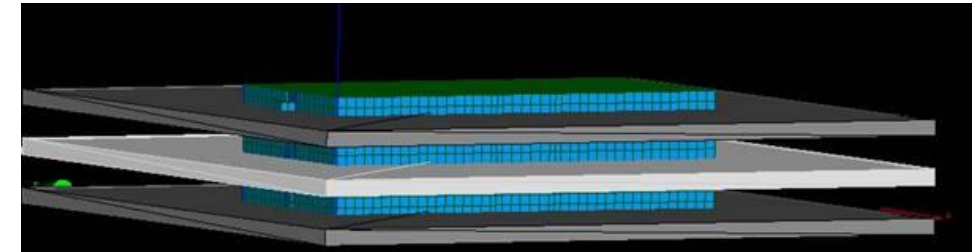
- Specific development for this study

- Python library to drive global model within Persalys _exec (3600 lines)
 - Automatic meshing generation (to account uncertain geometric parameter)
 - Run Cases (Saturne and TAeZoSypro) in parallel
 - ❑ Handle RESU directory unique name
 - ❑ Possibility to restart 3D model if convergency not reached
 - ❑ Possibility to only read already run case to extract other variable of interest
 - Post processing tools (matplotlib graph generation)

- Features to monitor global quantities for code saturne

- Mean Profiles on all or part of the mesh (MEDCoupling or CS STL) by layer (c++ 3800 lines)
- Balance by zone (surfaces/volumes) to monitor radiant/convective thermal exchanges and hvac power (3000 c lines)

- Various Python UNIX tools to handle amount of data generated (1 To of data generated for a total of 12 000 cpus day)
- _exec function writing for models drive (5800 lines)
- Python meshing function



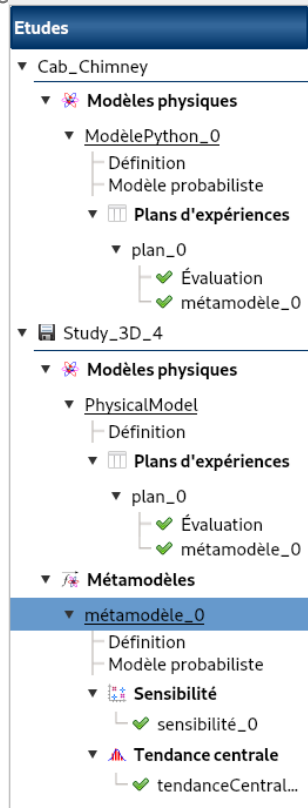
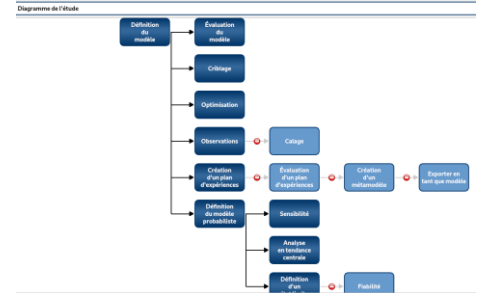
3D numerical experiments – Workflow

- Setup meshing script (uncertain geometrical parameter parts)
- Setup Code Saturne (uncertain parameter : physical properties / limite conditions)
- Setup equivalent 0D/1D TAeZoSysPro model (with accounted uncertain parameters)
- Setup persalys function `_exec` and test (if `__name__=='__main__':`)
- Generate Persalys study with several decoupled steps
 - Use of Persalys YACS model modified with the use of slurm study feature : enable executables launch with `srtn` command within an allocated batch on cluster (default YACS model launched branches with `srtn` command, preventing other use of `srtn` command)
 - Example : evaluate 300 point, each point require 20 Cpus, by block of 10 (total of 6 GAIA nodes with 204 Cpus requested)
 - Generate probabilistic experience plan based on uncertain parameters probabilistic model and csv export
 - Create another study with the same YACS model and create and imported csv experience plan
 - Run the experience plan as many time as required for all points to reach convergency
 - Create other YACS model, same uncertain parameters but with other variables of interest
 - `_exec` function will only read already run point
 - 1 for power balance / 1 for temperatures / 1 for dimensionless number

Result analysis within Persalys

- Experience plan result analyse
- Use experience plan as data model to generate MetaModel
 - Perform sensivity/central tendency analyses
 - Export MetaModel to improve 0D/1D models

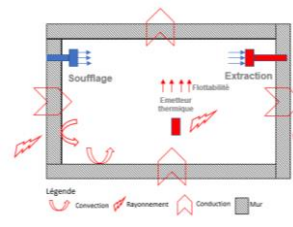
```
import salome_ot
import os
study,ModèleYACS_0=salome_ot.getYacsPyStudy(code)
import pydefx
pydefx_path = os.path.dirname(pydefx.__file__)
light_executor_path = os.path.join(pydefx_path, "plugins", "lightexecutor.py")
mybuilder = pydefx.slurmbuilder.SlurmBuilder(executor=light_executor_path)
myModel = pydefx.SlurmStudy(schemaBuilder=mybuilder)
ModèleYACS_0.setJobModel(myModel)
```



Phénomènes physiques identifiés

• Les principaux phénomènes physique identifiés à l'échelle du local industriel sont:

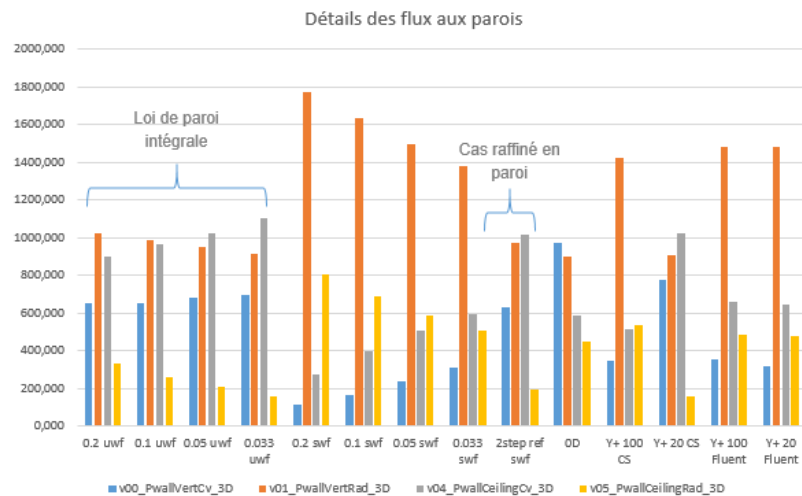
Phénomènes physiques	Paramètres géométriques	Paramètres matériaux	Conditions aux limites	Nombre adimensionnel
Conduction thermique dans les solides Béton - Emetteur	Epaisseur des parois Surface des parois	Conductivité thermique c_p et ρ	Température adjacente Flux adjacent	Biot
Convection naturelle murs	Surface, Longueur caractéristique	Viscosité ρ, λ_{fluide}	Température adjacente Flux adjacent	Rayleigh Prandtl
Rayonnement Mur – Mur Emetteur – Enveloppe Enveloppe -Mur	Surface Angles solides entre surfaces	Emissivité	Flux solaire Température ciel	-
Inertie thermique	Géométrie des masses Surface d'échange	ρ, c_p, V, λ	-	Fourrier
Transferts enthalpiques	Géométrie du local Position des bouches de soufflages/extraction	$\beta, \rho, c_p, \mu, \lambda$	Débit ventilation Localisation entrée/sorties Température de soufflage Charge thermique	-
Stratification thermique	Géométrie du local Géométrie source thermiques (incluant leurs freins aérauliques)	$\beta, \rho, c_p, \mu, \lambda$	Température de soufflage Vitesse de soufflage Charge thermique	Richardson
Effets de jets	Géométrie des obstacles	$\beta, \rho, c_p, \mu, \lambda$	Vitesse de soufflage	Reynolds Richardson
Panaches thermiques	Géométrie du local Géométrie des sources	$\beta, \rho, c_p, \mu, \lambda$	Charge thermique	Rayleigh Reynolds
Configuration Rayleigh Bénard	Géométrie du local	$\beta, \rho, c_p, \mu, \lambda$	Température adjacente	Rayleigh
Mélange turbulent	Géométrie de la pièce Nature des obstacles	$\beta, \rho, c_p, \mu, \lambda$	Vitesse soufflage Charge thermique	Reynolds Rayleigh



Expérience numériques 3D

Configuration

- Utilisation de loi de paroi thermique intégrale pour une estimation correcte des flux thermiques (précision l'ordre de 10-20 %)
- Couplage de l'ensemble des phénomènes (conduction thermique solide, échanges conducto/convectifs, échanges radiatifs) étant donné l'importance de chacun d'entre eux sur la distribution du champs de température
- Utilisation de maillages hexaédriques par bloc avec une taille de cellule élémentaire de 0,01 m
- Modèle de turbulence : $k-\omega$ SST (faible sensibilité au modèle de turbulence vis-à-vis des grandeurs d'intérêt intégrales identifiées)
- Approche 3D zonale ou CFD macroscopique : La résolution de la structure local fine de l'écoulement n'est pas réalisée.



Configuration numérique

- Schémas numériques
 - Upwind pour la turbulence
 - SOLU pour les autres variables
- Précision solver : 10^{-5} sauf la température 10^{-6}
- Paramètres temps
 - Variable en temps (IDTVAR=1)
 - ❑ Ref time step : 0,1 s
 - ❑ Time step maximal variation : 0,01
 - CDTVAR :
 - ❑ 20 pour la température pendant 1000-1500 itérations
 - ❑ Passage à 1 via cs_user_extra_operations.c au bout de 1000-1500 itérations, calcul poursuivi jusqu'à 3000- 4000 itérations
- A affiner selon les cas

Post Traitement

- Export de tous les bilans et profils en csv
- Export vtk de l'état final
- Export Catalyst : Vue globale, coupe en température et en vitesse
- Génération d'un rapport LaTeX globale : 2 pages par cas d'évaluation