

SMART CFD PROXY

REPLICATING CFD RESULTS ACCURATELY & AT HIGH SPEED

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Outline

- Smart Proxy
 - Engineering Application of Big Data Analytics
- Original Application in Reservoir Modeling
 - Tracking Pressure & Saturation Changes at the Grid Block Level
- Application to CFD
 - Multi-Phase Fluidized Bed (Proof of Concept)

Computational Science

- Numerical solution to complex, multi-physic, non-linear, partial differential equation emerged, when
 - Engineering models became too complex to be solved analytically,
 - High speed computers became widely available.

Computational Footprint

- Our models soon increased in size such that use of super computers and CPU clusters became a necessity to solve serious problems.
- Proxy models are used to address the extensive computational cost of the numerical simulation models.

Proxy Models

- Proxy models attempt to behave like the original models but with lower computational cost.
- The current paradigm for developing proxy models is to simplify the original model.
- Existing proxy models are either:
 - Statistics-based (response surfaces – curve fit at specific locations)
 - Physics-based (Reduced Physics)
 - Coarse Models (low resolution)

Paradigm Shift in Proxy Modeling

- In this new paradigm of developing proxy models,
 - The proxy model represents the entire system.
 - Physics of the model is **NOT** simplified.
 - Model resolution is **NOT** reduced.
- In this new paradigm, data (generated by CFD), and not the physics, will be the basis for the proxy model.

Paradigm Shift in Proxy Modeling

- Smart Proxy uses the pattern recognition capabilities of Artificial Intelligence and Data Mining (Big Data Analytics) to **LEARN** and then **MIMIC** the behavior of the original CFD.
- Smart Proxy is trained using massive amount of data that is generated from (only a handful of) CFD models runs and validated using blind CFD runs.

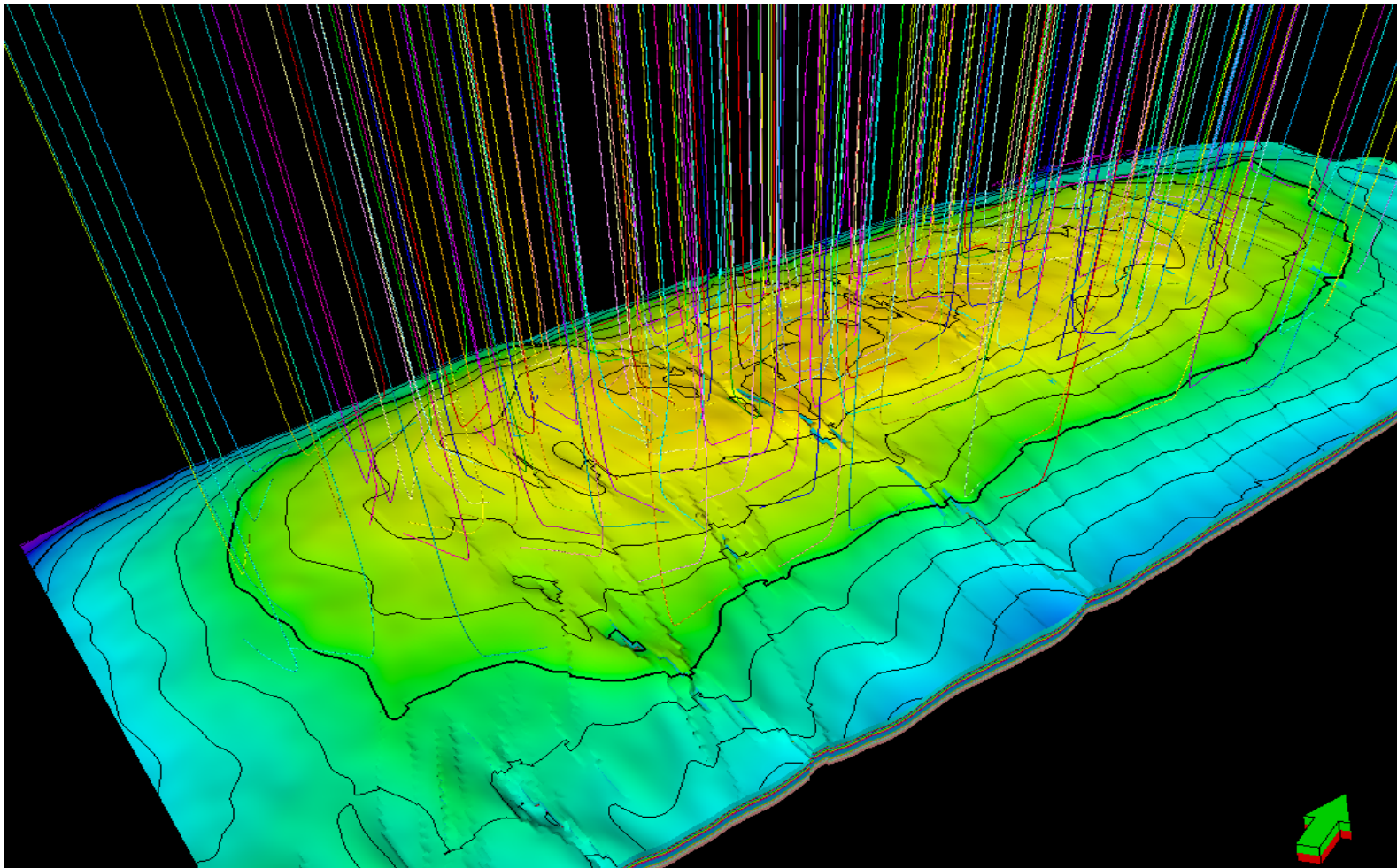
Paradigm Shift in Proxy Modeling

- A single run of the original model is treated as a treasure trove of data and information that can be used to train and calibrate a new, smart model.
- This technology is the manifestation of **“BIG DATA ANALYTICS”** in engineering problem solving, specifically, in the numerical modeling arena.

Surrogate Reservoir Modeling - SRM

SMART PROXY MODEL FOR HIGH SPEED REPLICATION OF THE RESULTS OF NUMERICAL RESERVOIR SIMULATOR

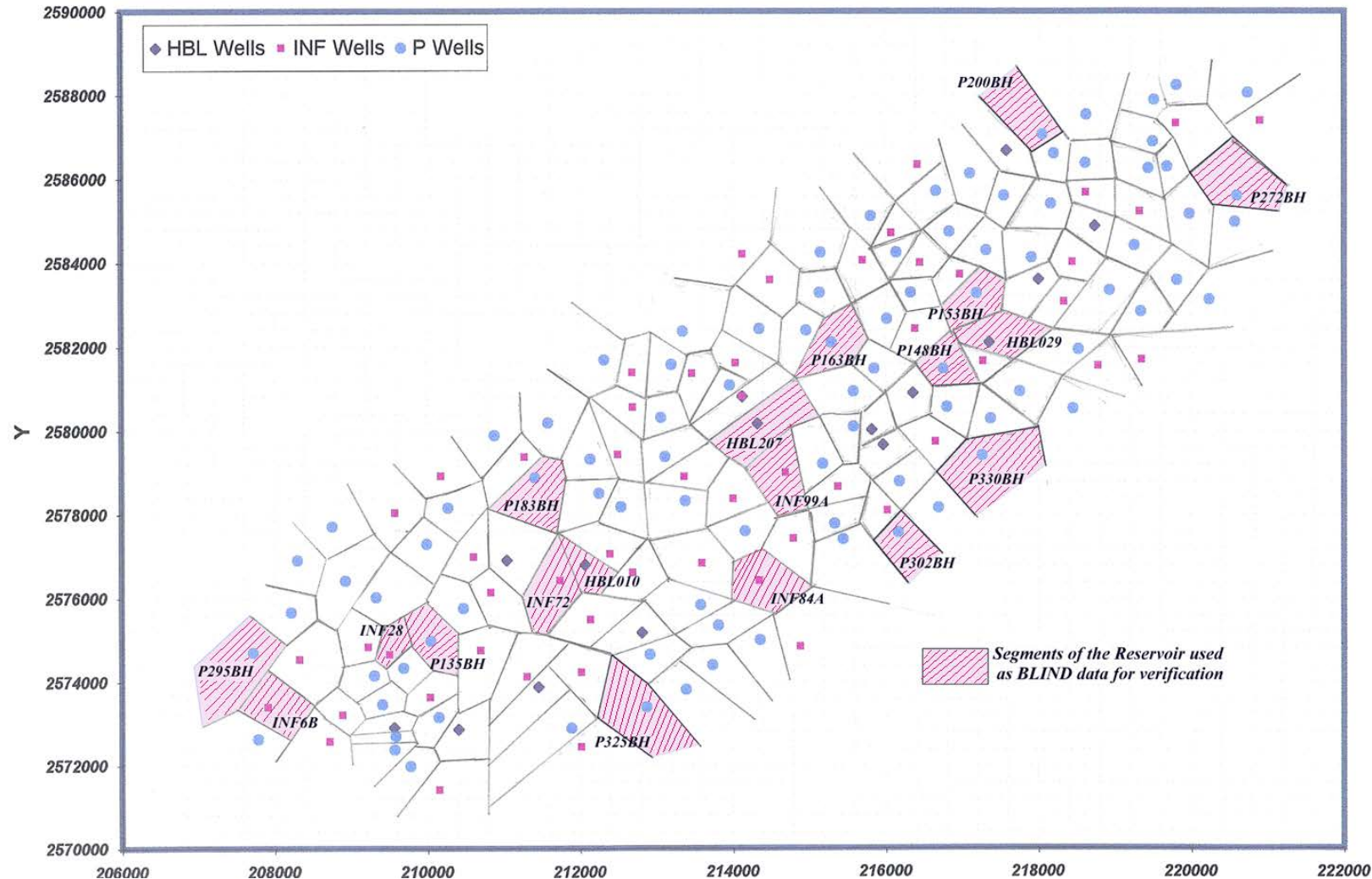
Numerical Model Characteristics



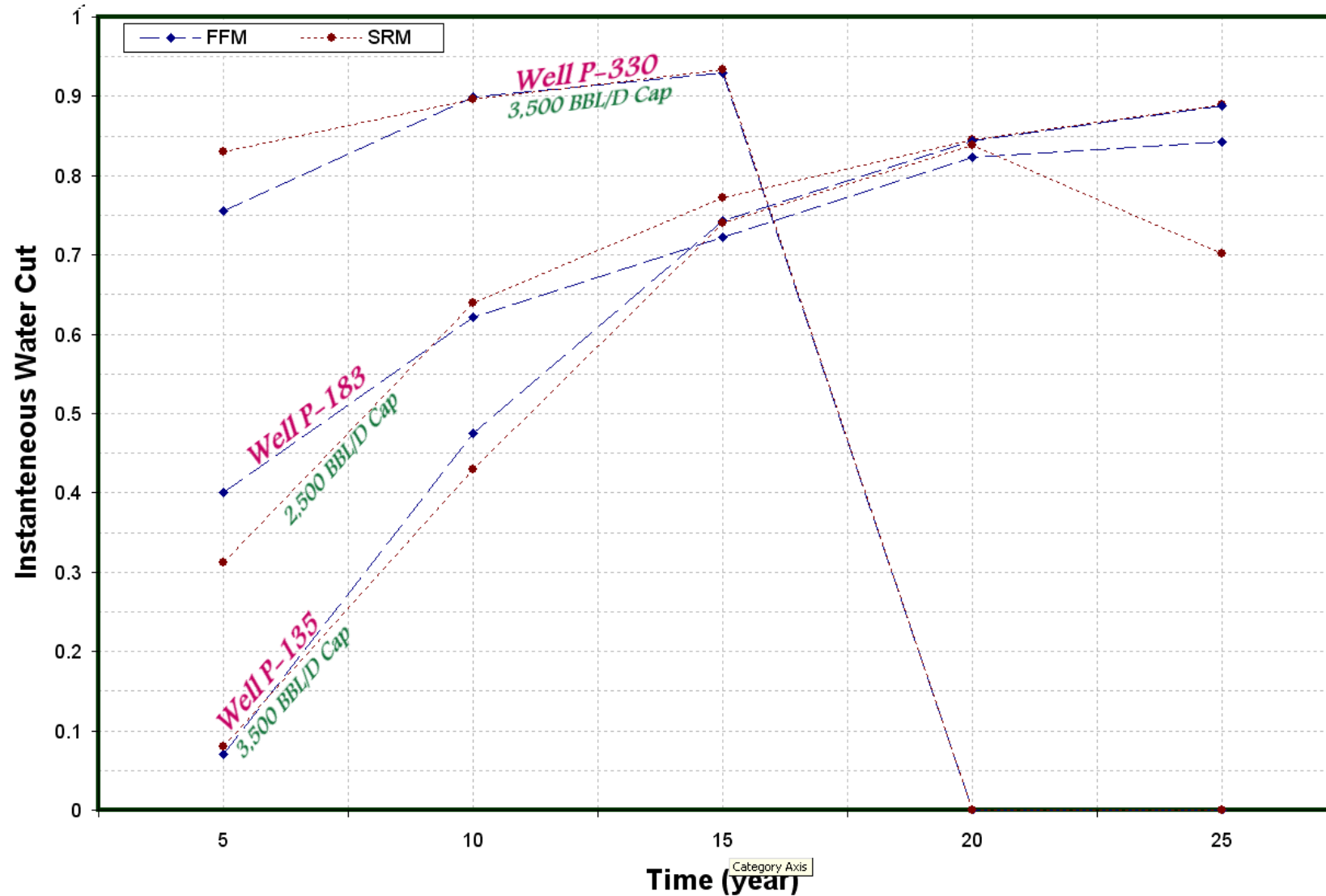
Study Completed	2005
Implemented	2006
Field Results Published (SPE 170664)	2014

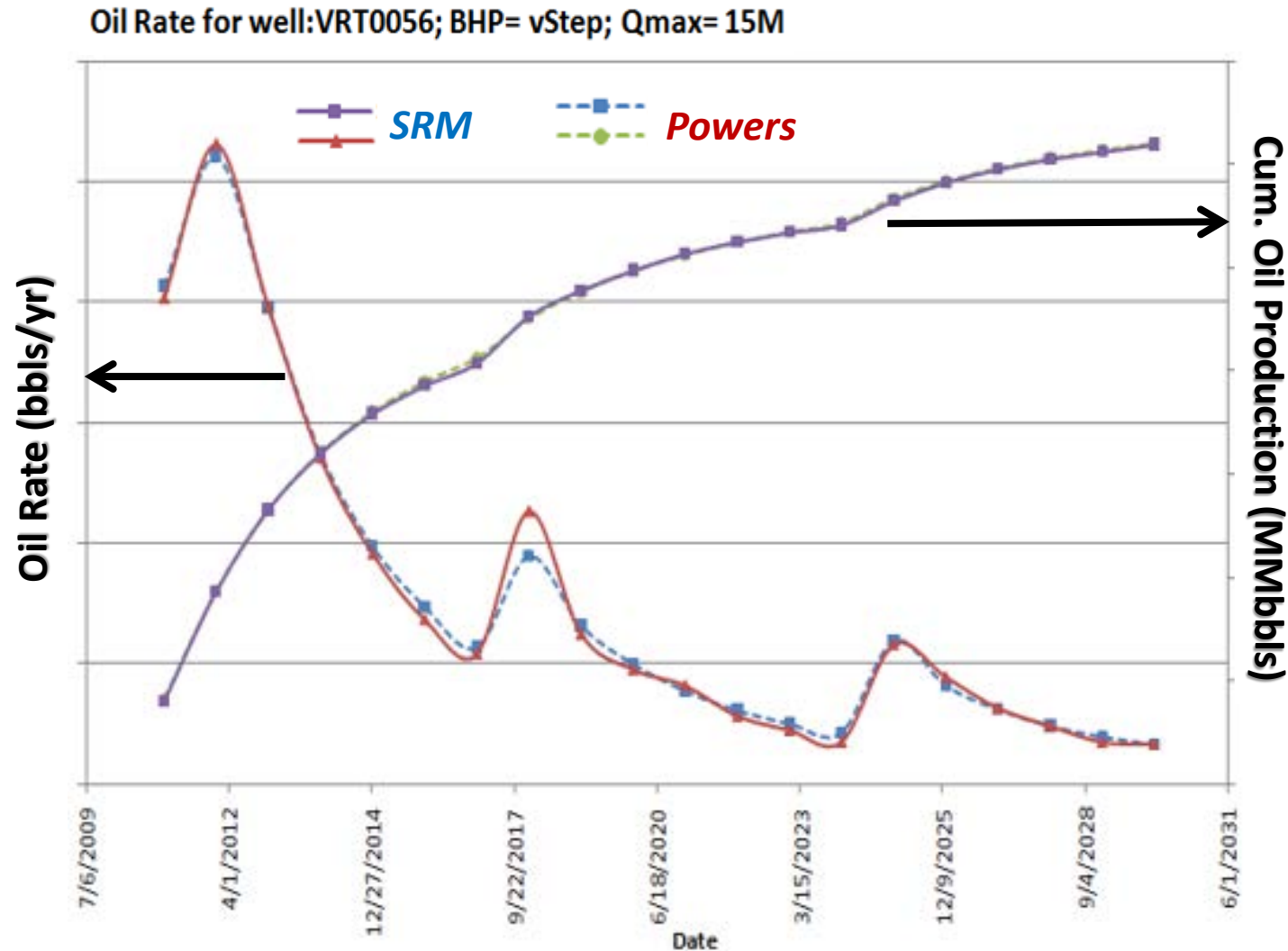
Size of the Full Field Model	1MM Grid Blocks
Formation Type	Naturally Fractured Carbonate
Number of Wells	167 (Horizontal)
Simulation Run	10 Hours on 12 parallel CPU

Field Production Optimization Study



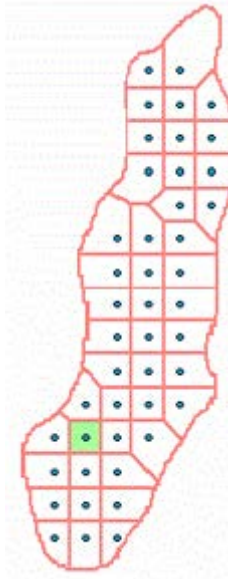
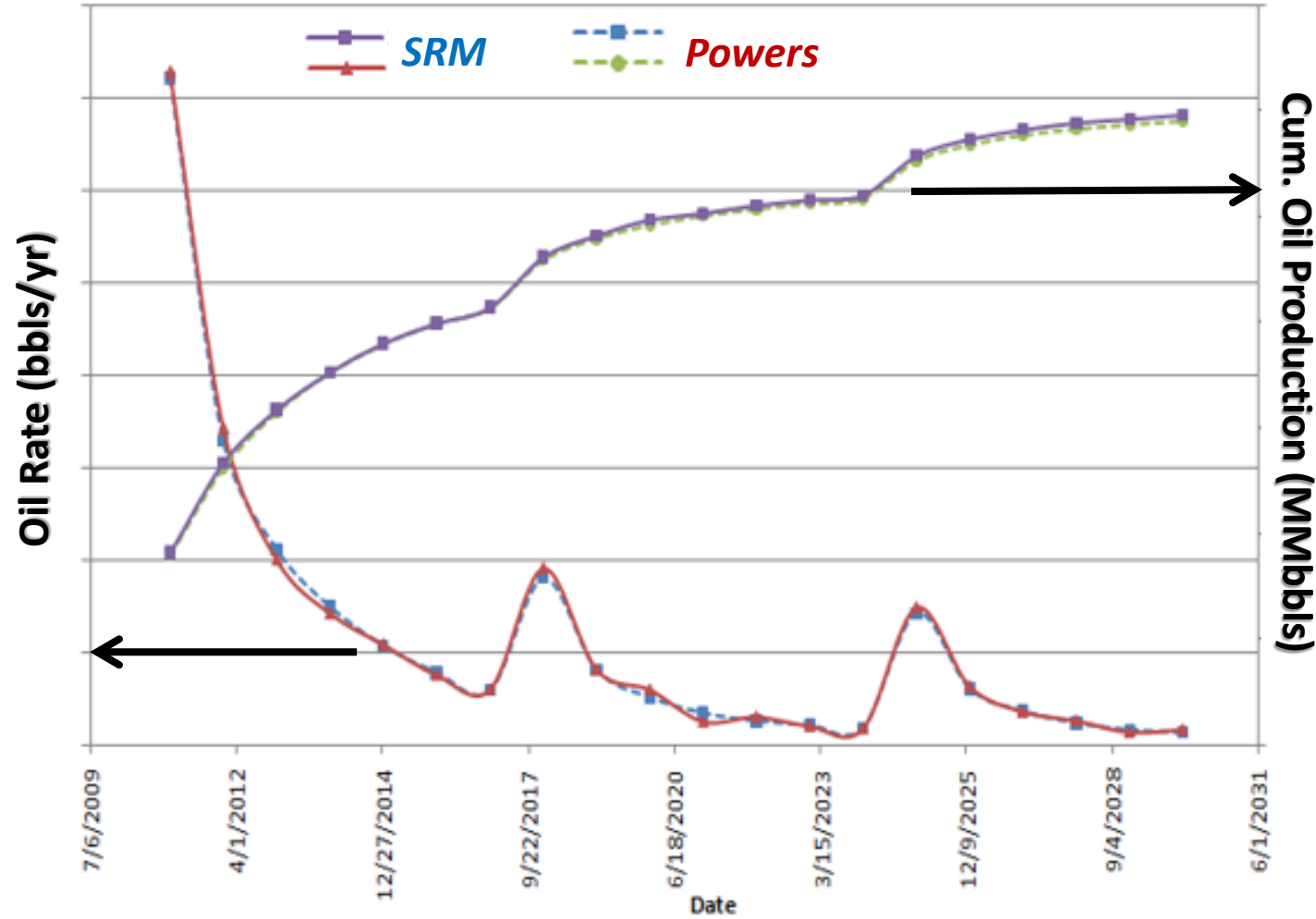
Field Production Optimization Study





*Black Oil Simulation
Green Field
6.5 MM Grid Blocks*

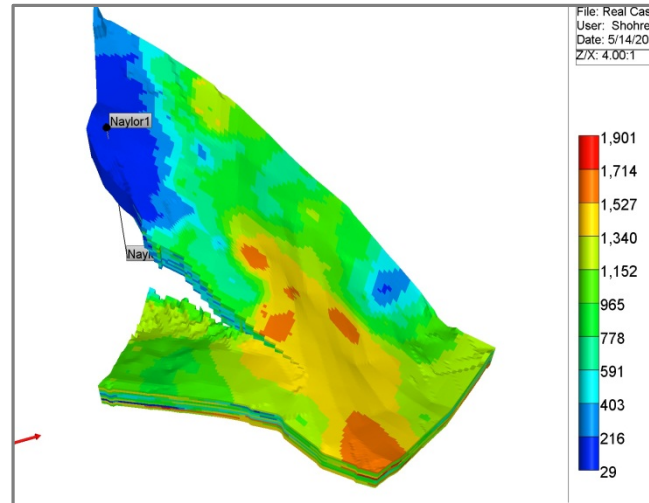
Oil Rate for well:VRT0019; BHP= vStep; Qmax= 15M



*Black Oil Simulation
Green Field
6.5 MM Grid Blocks*

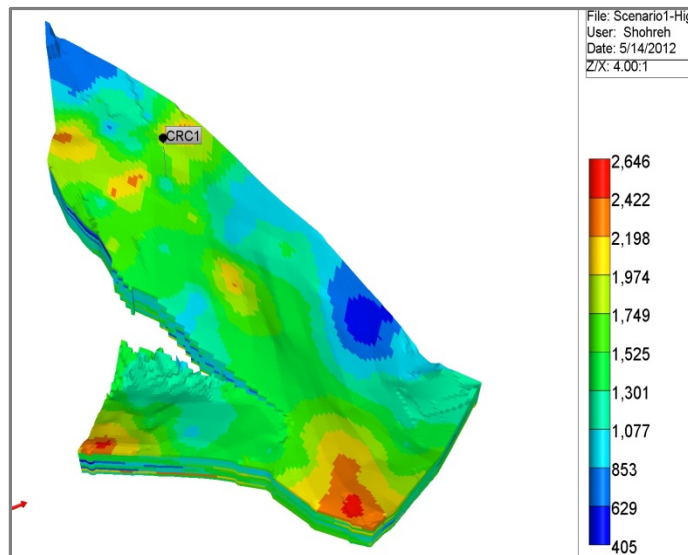
Otway CO₂ Seq. Project, Australia

History Matched Perm

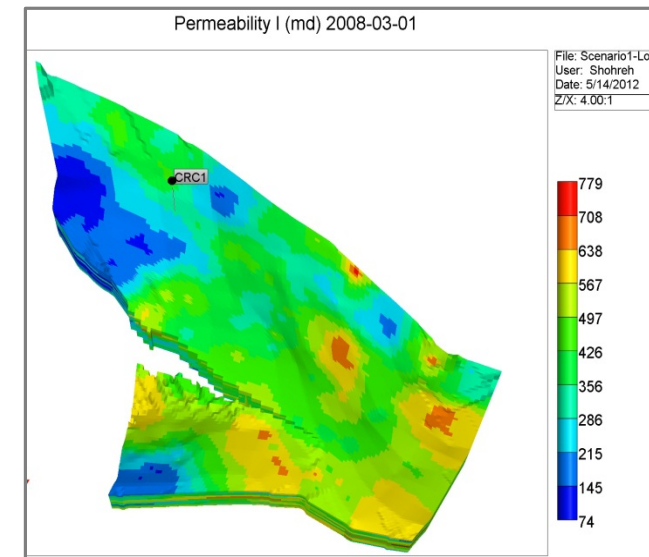


Permeability Maps

High perm

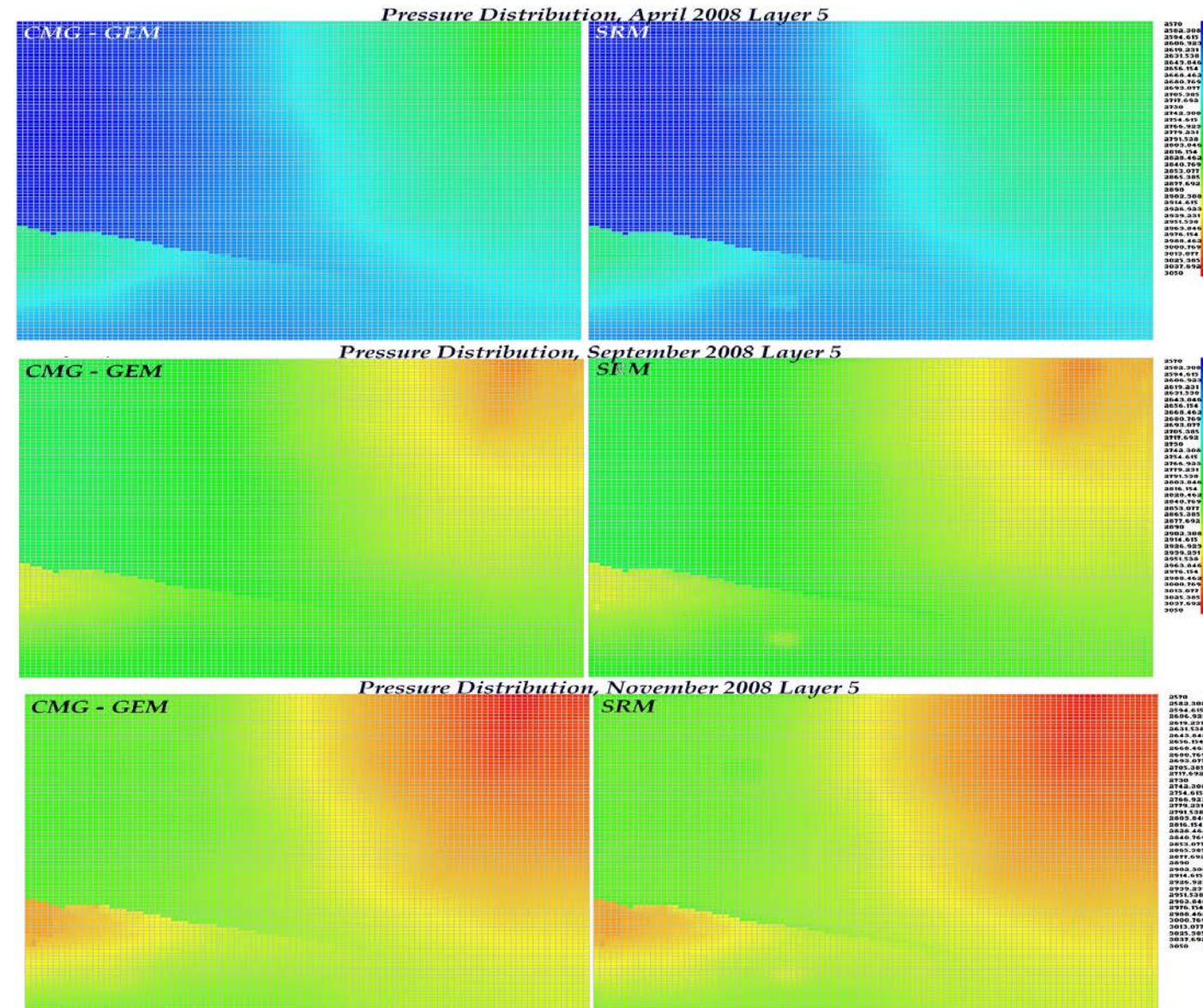


Low perm



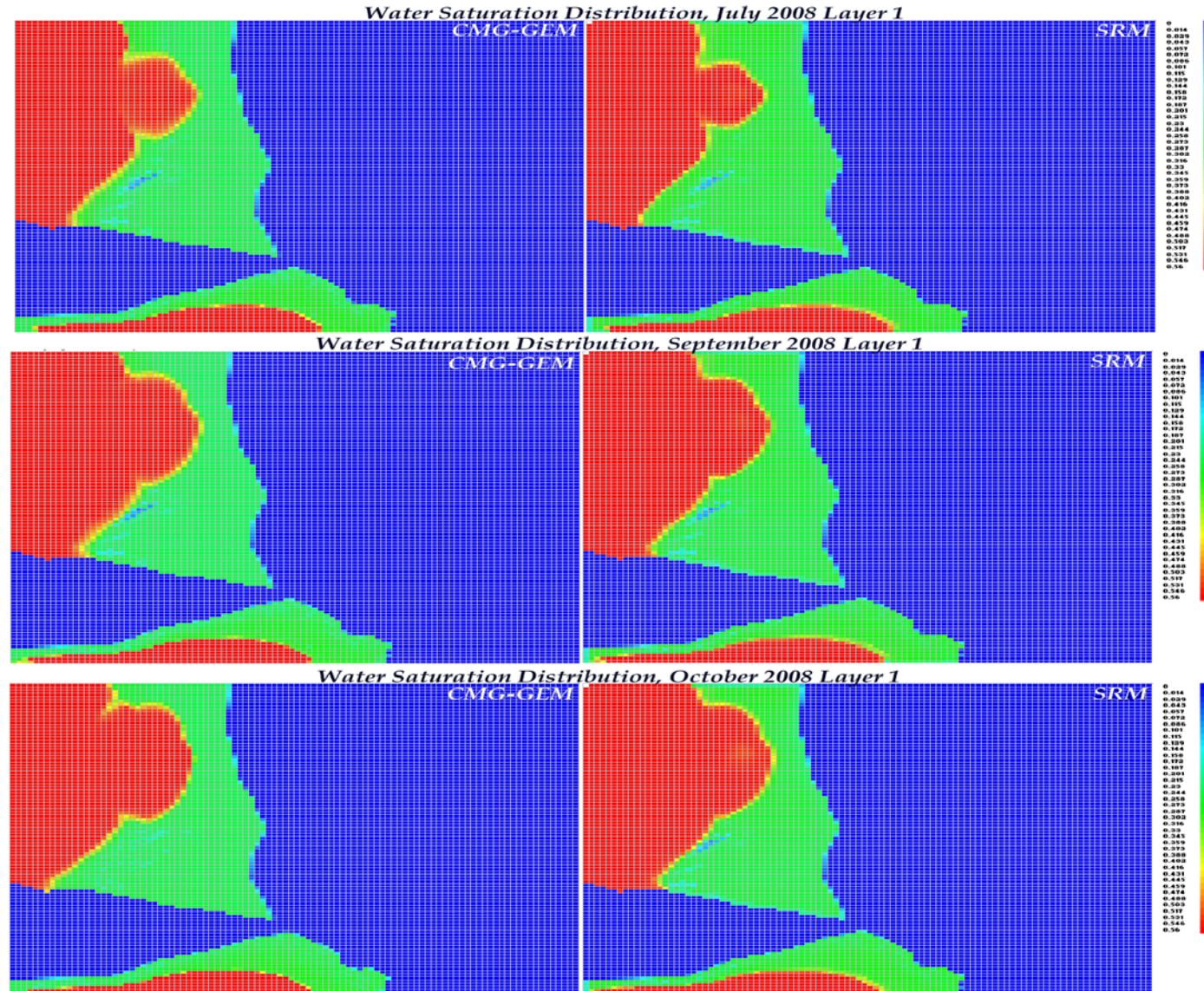
Otway CO₂ Seq. Project, Australia

Pressure Distribution



Otway CO₂ Seq. Project, Australia

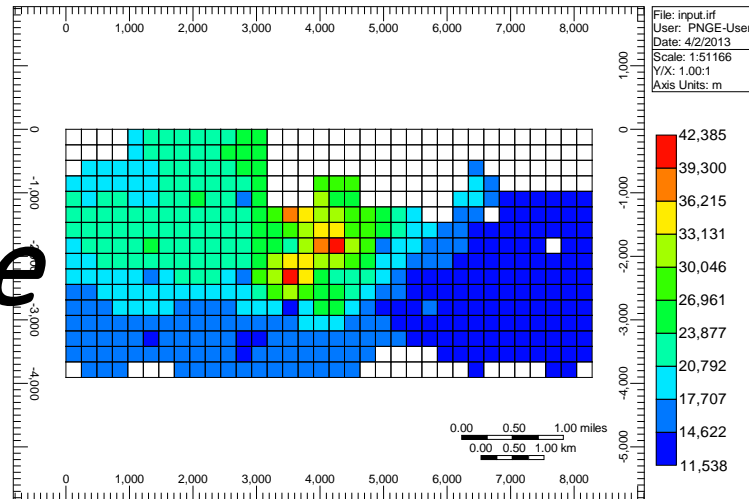
Water Saturation Distribution



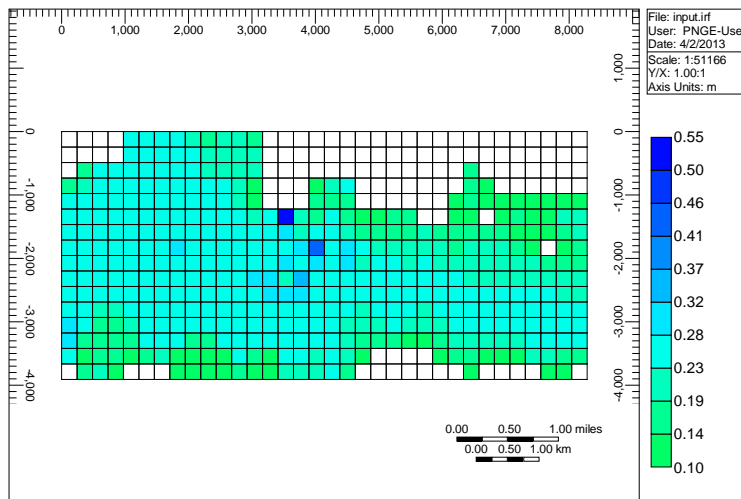
SACROC CO₂ EOR Project, Texas

The Coarse Reservoir Model

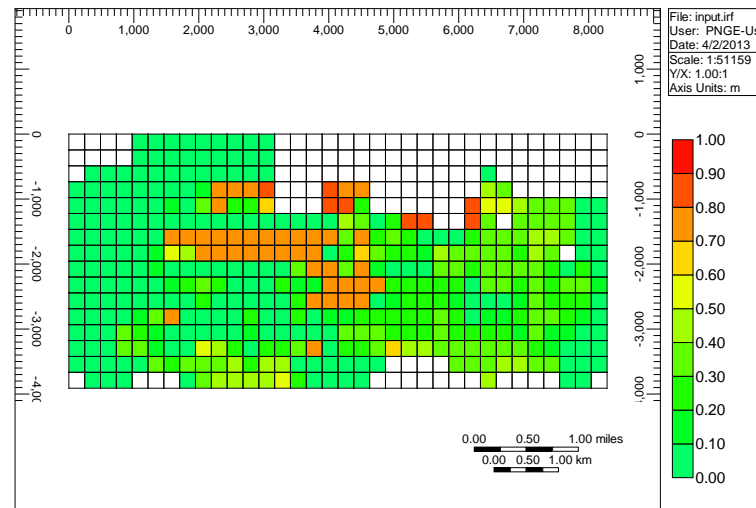
Pressure (kPa) 2172-01-01 K layer: 18



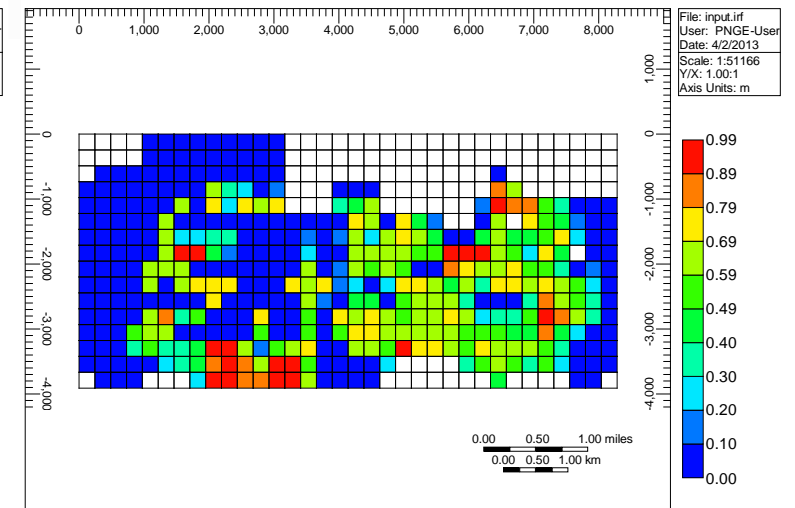
Water Saturation 2172-01-01 K layer: 18



Gas Saturation 2172-01-01 K layer: 18



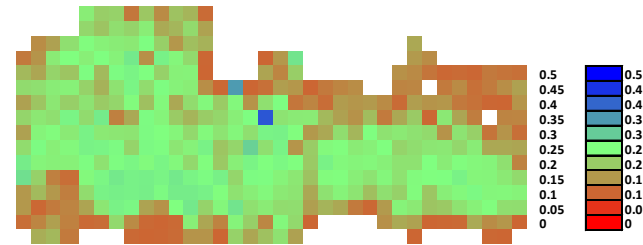
Gas Mole Fraction(CO2) 2172-01-01 K layer: 18



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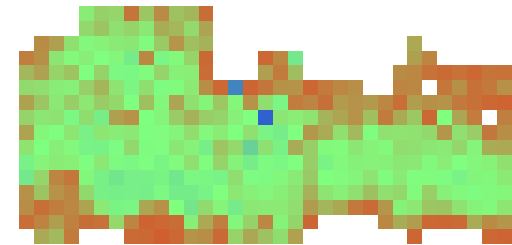
Water Saturation Distribution

Simulator (CMG) Results



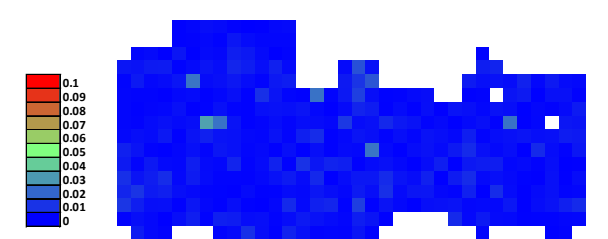
Actual Data: Realization # 1, Layer= 18, 100 years after Injection, Feature= SW %

SRM Results

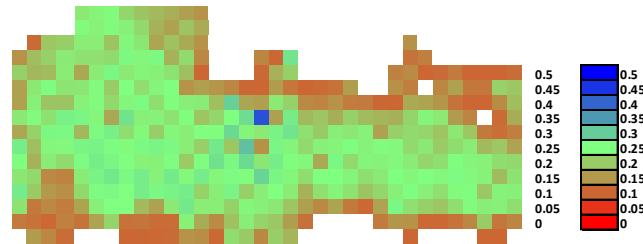


SRM Data: Realization # 1, Layer= 18, 100 years after Injection, Feature= SW %

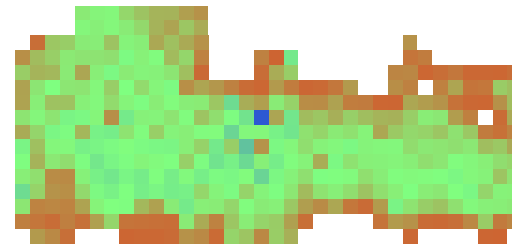
% Relative Error



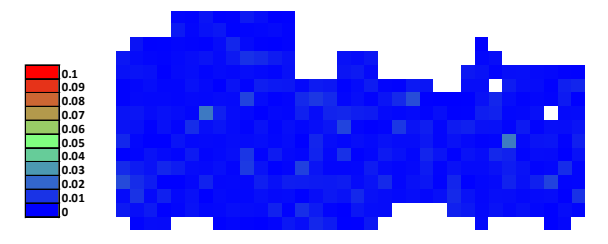
Realization # 1, Layer= 18, 100 years after Injection, Feature= Absolute Error of SW %



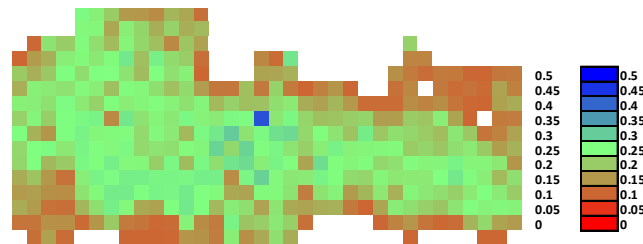
Actual Data: Realization # 11, Layer= 18, 100 years after Injection, Feature= SW %



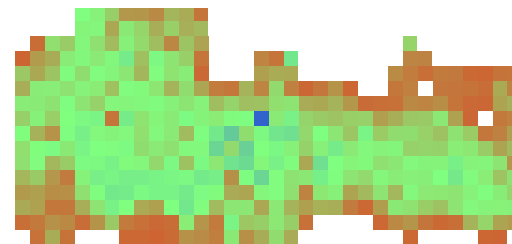
SRM Data: Realization # 11, Layer= 18, 100 years after Injection, Feature= SW %



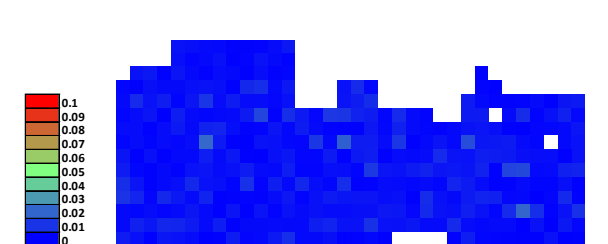
Realization # 11, Layer= 18, 100 years after Injection, Feature= Absolute Error of SW %



Actual Data: Realization # 12, Layer= 18, 100 years after Injection, Feature= SW %



SRM Data: Realization # 12, Layer= 18, 100 years after Injection, Feature= SW %



Realization # 12, Layer= 18, 100 years after Injection, Feature= Absolute Error of SW %

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CO₂ Mole Fraction Distribution

Simulator (CMG) Results



Actual Data: Realization # 8, Layer= 18, 100 yrs after Injection,
Feature= Pressure (psi)

SRM Results



SRM Data: Realization # 8, Layer= 18, 100 yrs after Injection,
Feature= Pressure (psi)

% Relative Error



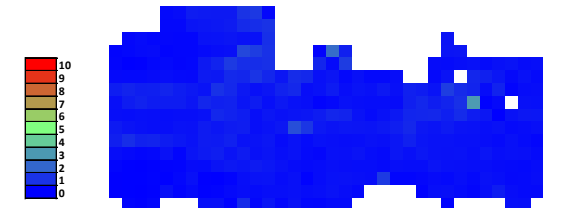
Realization # 8, Layer= 18, 100 yrs after Injection, Feature=
Relative Error %



Actual Data: Realization # 9, Layer= 18, 100 yrs after Injection,
Feature= Pressure (psi)



SRM Data: Realization # 9, Layer= 18, 100 yrs after Injection,
Feature= Pressure (psi)



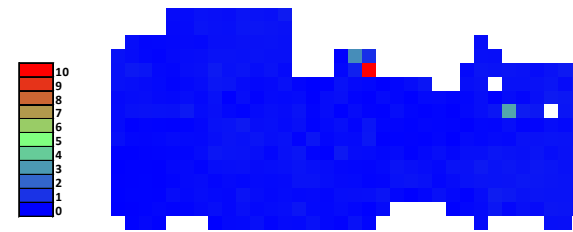
Realization # 9, Layer= 18, 100 yrs after Injection, Feature=
Relative Error %



Actual Data: Realization # 10, Layer= 18, 100 yrs after Injection,
Feature= Pressure (psi)



SRM Data: Realization # 10, Layer= 18, 100 yrs after Injection,
Feature= Pressure (psi)

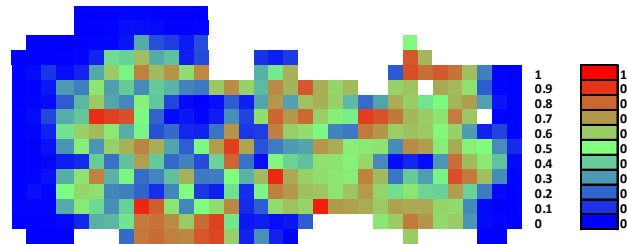


Realization # 10, Layer= 18, 100 yrs after Injection, Feature=
Relative Error %

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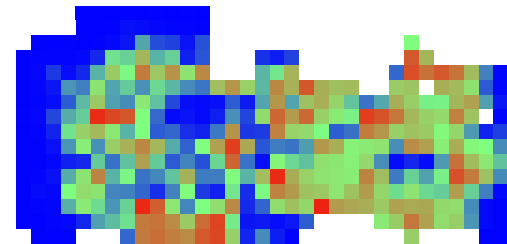
Pressure Distribution

Simulator (CMG) Results



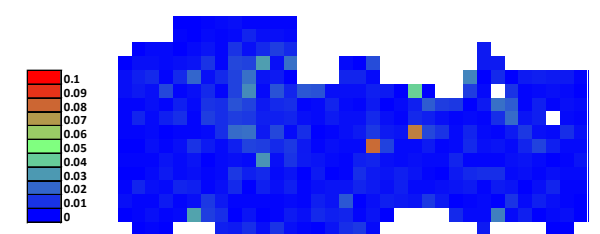
Actual Data: Realization # 9, Layer= 18, 9 years after Injection, Feature= CO2 Mole Fraction %

SRM Results

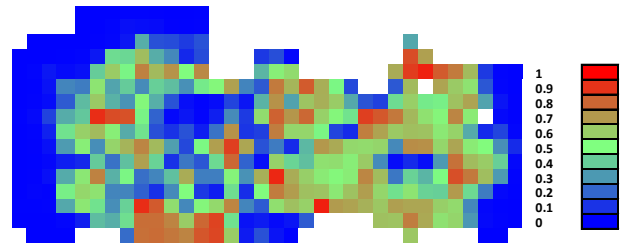


SRM Data: Realization # 9, Layer= 18, 9 years after Injection, Feature= CO2 Mole Fraction %

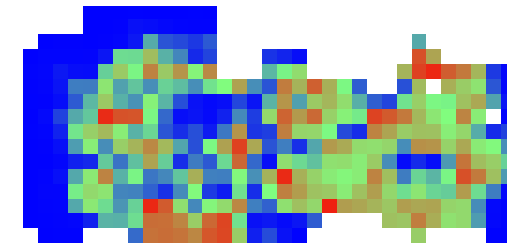
Absolute Error



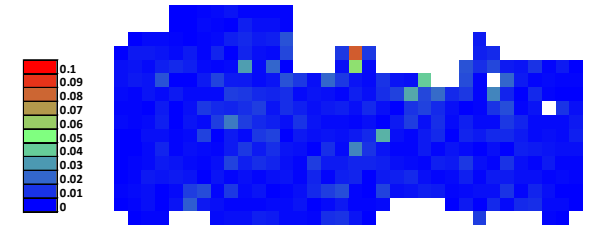
Realization # 9, Layer= 18, 9 years after Injection, Feature= Absolute Error %



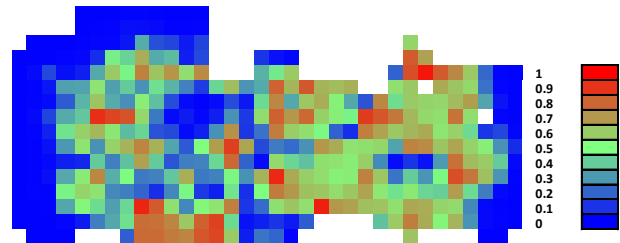
Actual Data: Realization # 10, Layer= 18, 9 years after Injection, Feature= CO2 Mole Fraction %



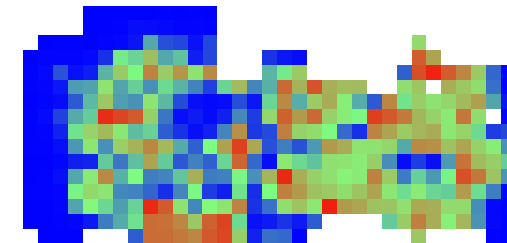
SRM Data: Realization # 10, Layer= 18, 9 years after Injection, Feature= CO2 Mole Fraction %



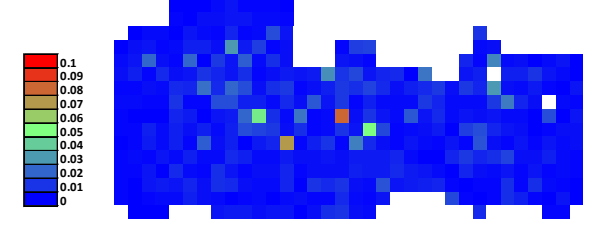
Realization # 10, Layer= 18, 100 years after Injection, Feature= Absolute Error %



Actual Data: Realization # 7, Layer= 18, 9 years after Injection, Feature= CO2 Mole Fraction %

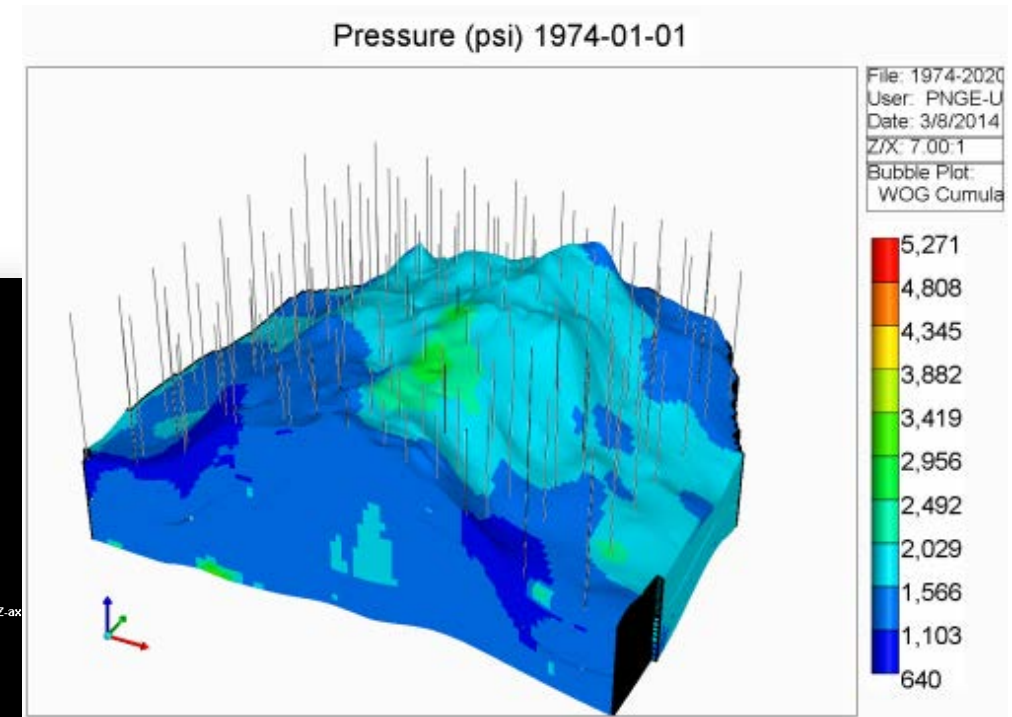
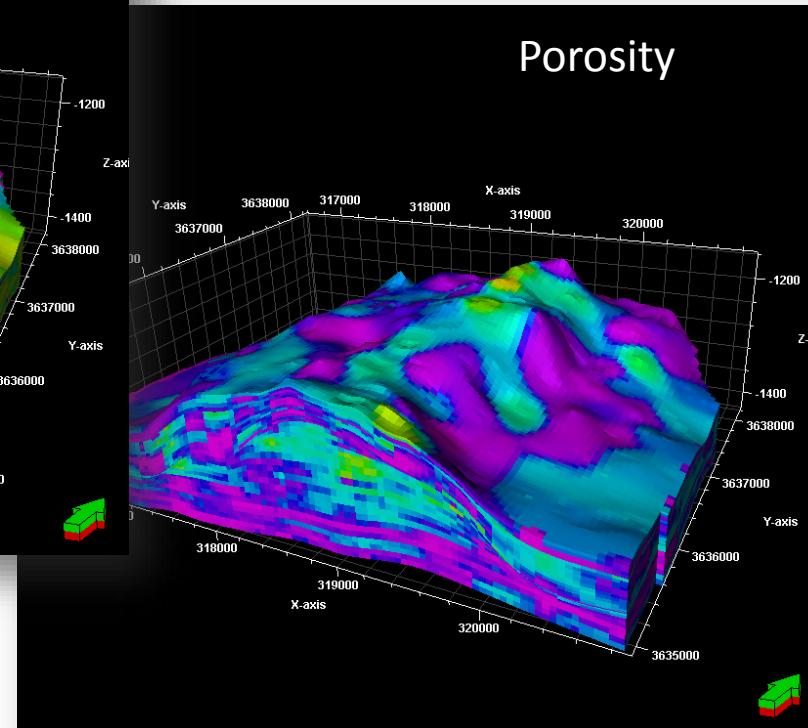
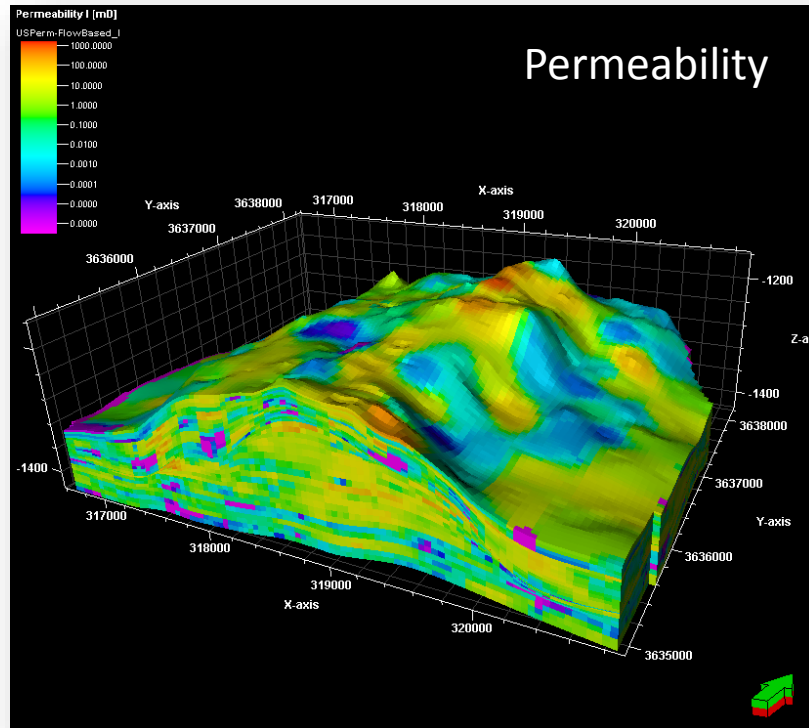


SRM Data: Realization # 7, Layer= 18, 9 years after Injection, Feature= CO2 Mole Fraction %



Realization # 7, Layer= 18, 100 years after Injection, Feature= Absolute Error %

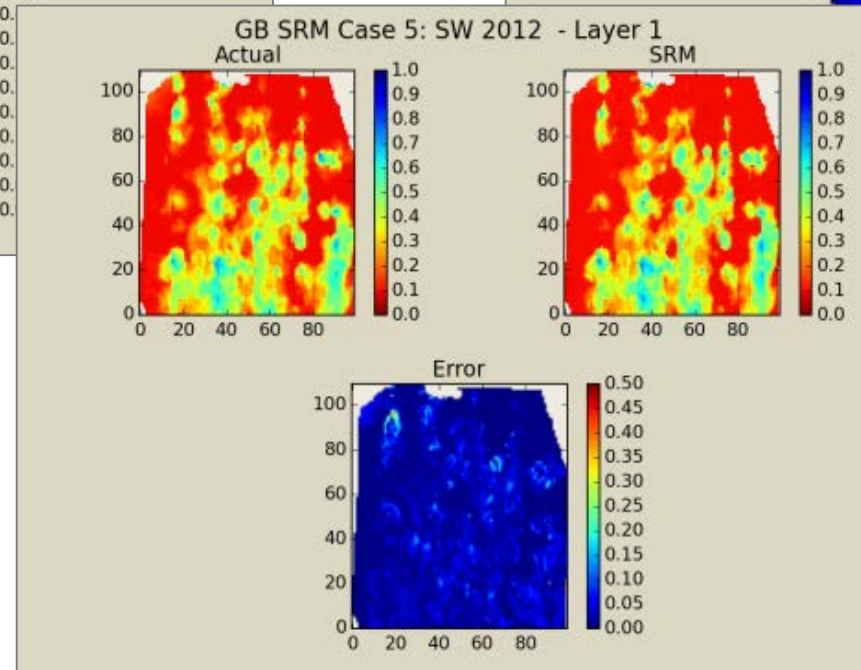
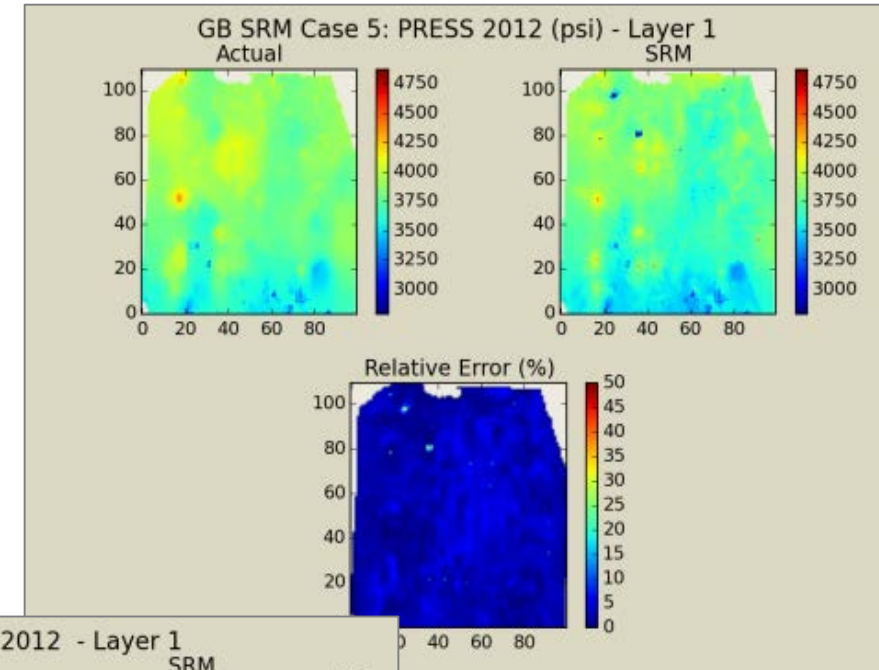
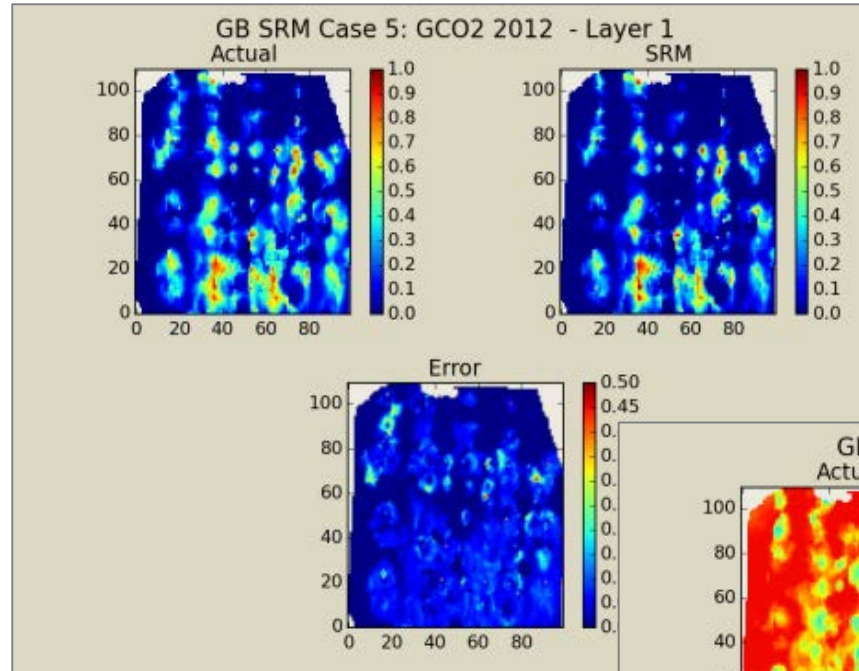
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High Resolution Model

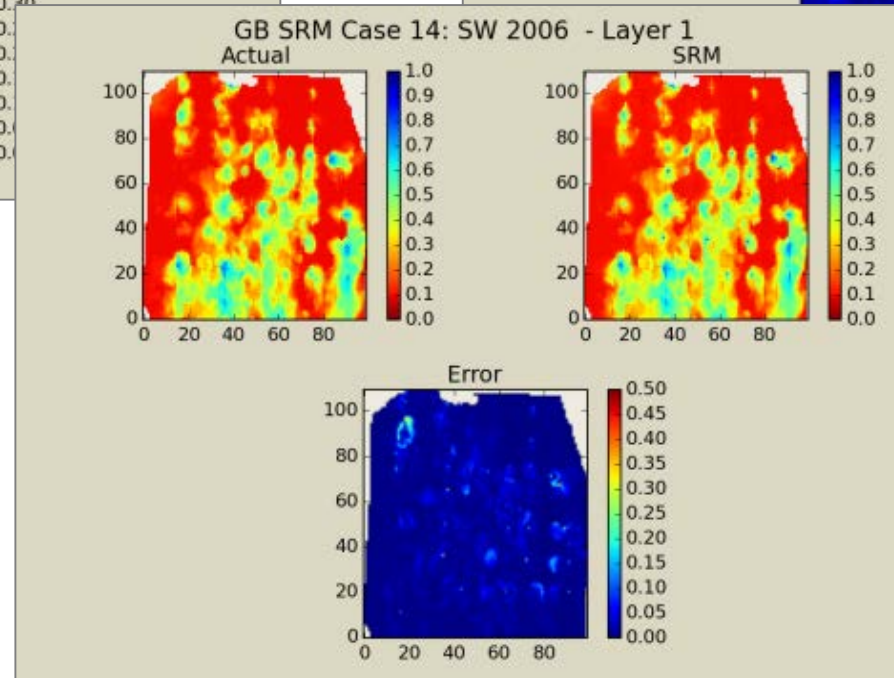
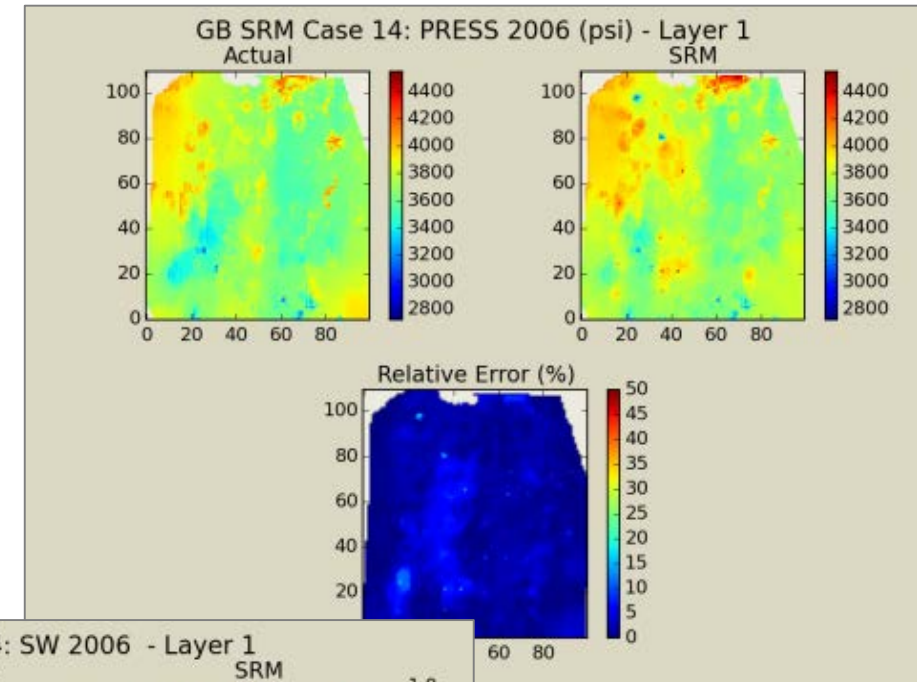
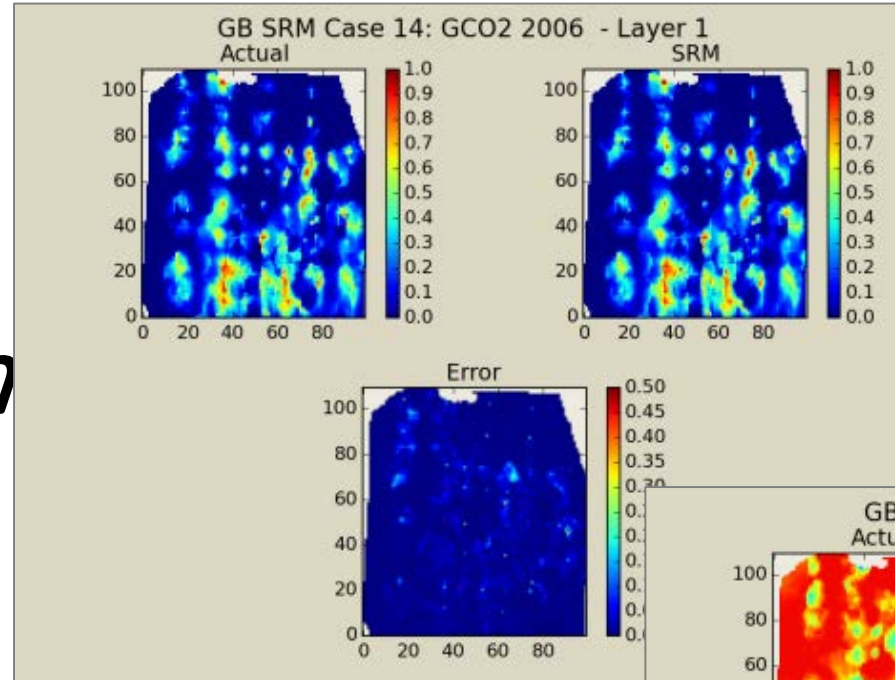
SACROC CO₂ EOR Project, Texas

*High
Resolution
Model*



SACROC CO₂ EOR Project, Texas

*High
Resolution
Model*



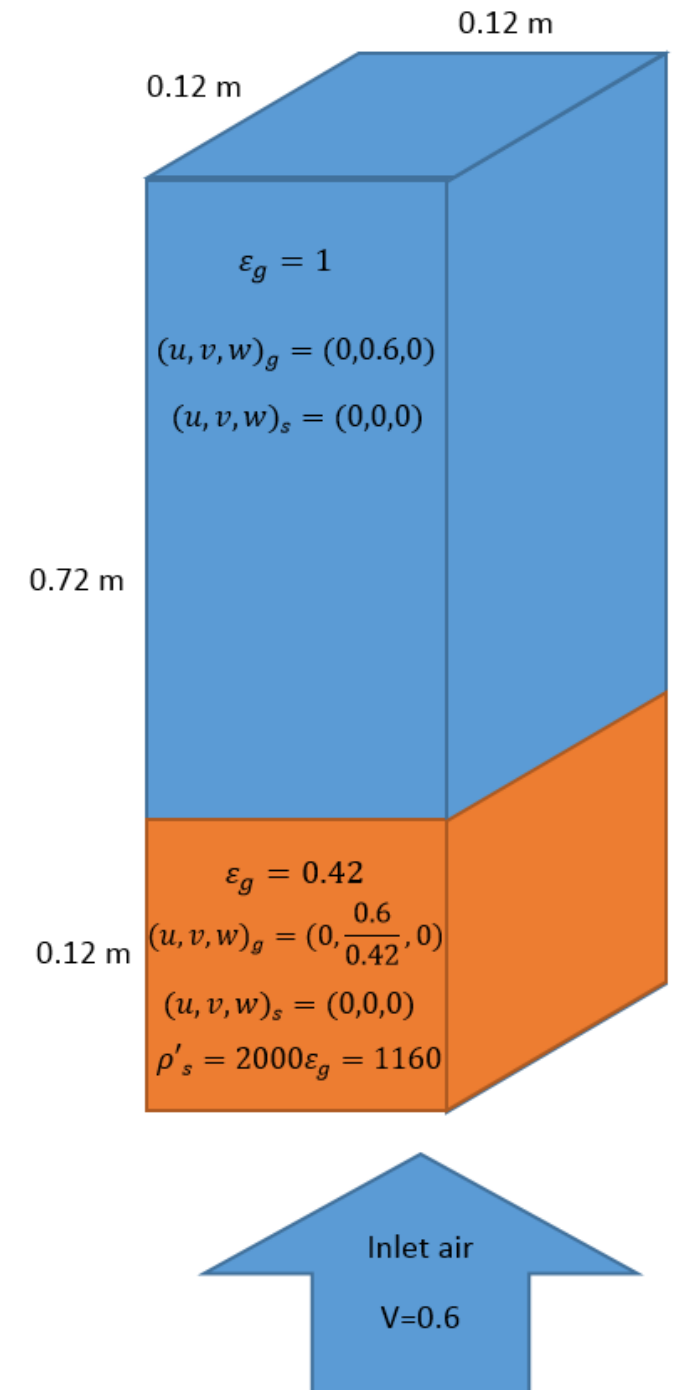
Smart Proxy for CFD

CAN SMART PROXY BE SUCCESSFULLY APPLIED TO CFD?

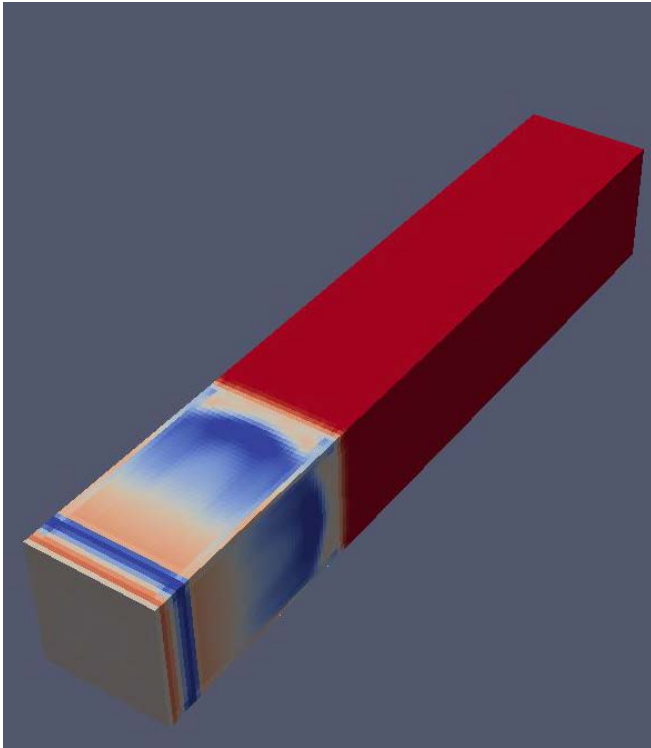
A Multi-phase CFD Model

- Initial pressure is 1 atm. in the entire domain
1 atm. = 101,325 Pa

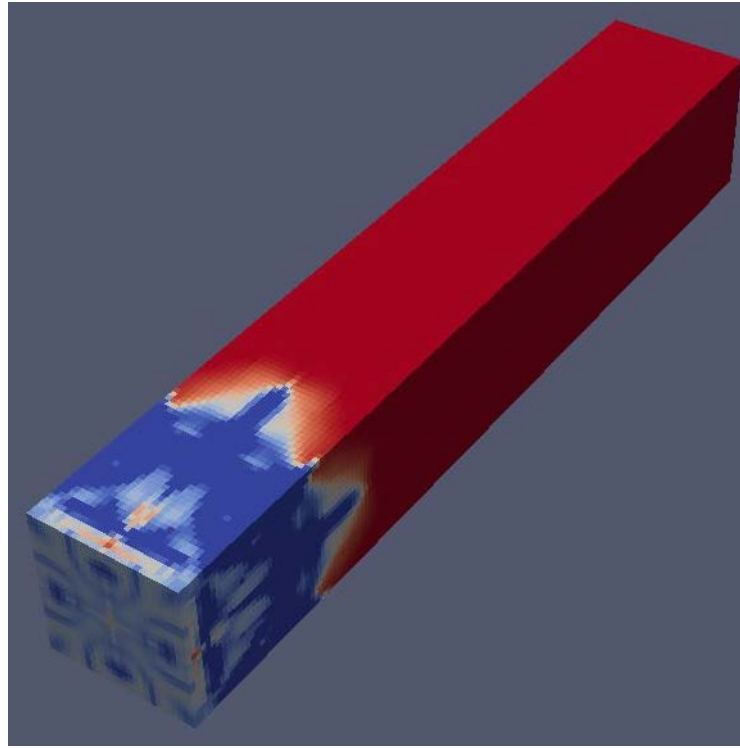
Grid Classification	I, j, k - Cell size	No. of Cell	No. of Grid
Coarse	8 x 48 x 8 (15 mm)	3,072	3,969
Medium	12 x 72 x 12 (10 mm)	10,368	12,337
Fine	18 x 108 x 18 (6.6 mm)	34,992	39,349
Very Fine	27 x 162 x 27 (4.4 mm)	118,098	127,792



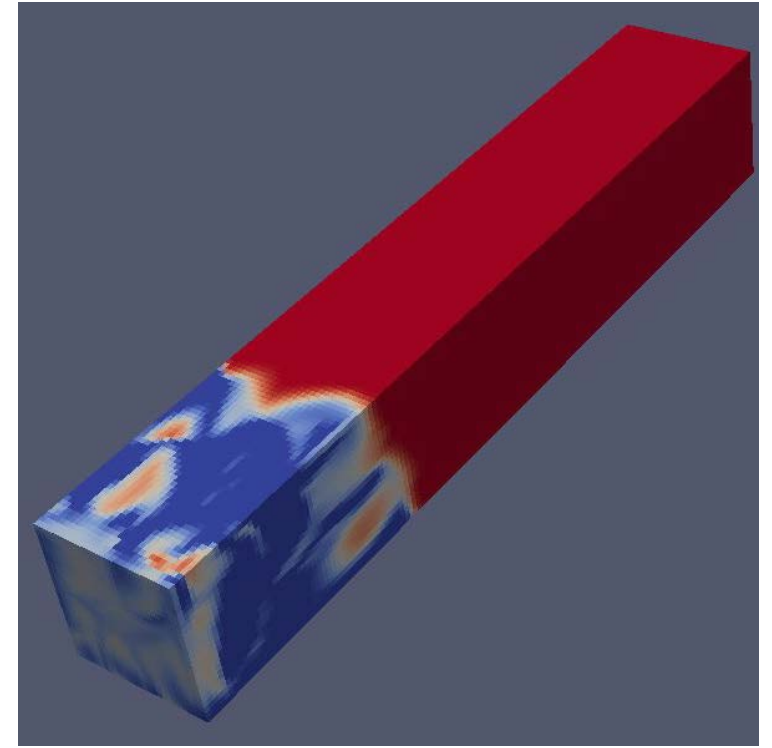
Three different physics



Early-time
Time step 200

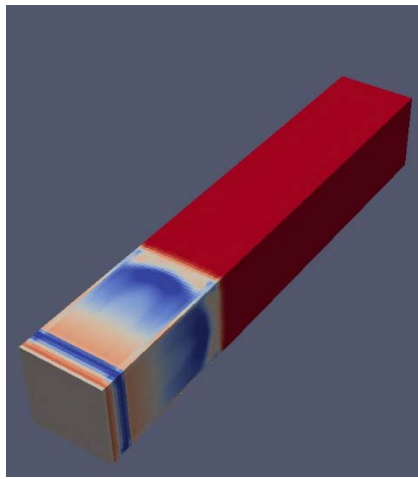


Mid-time
Time step 1000

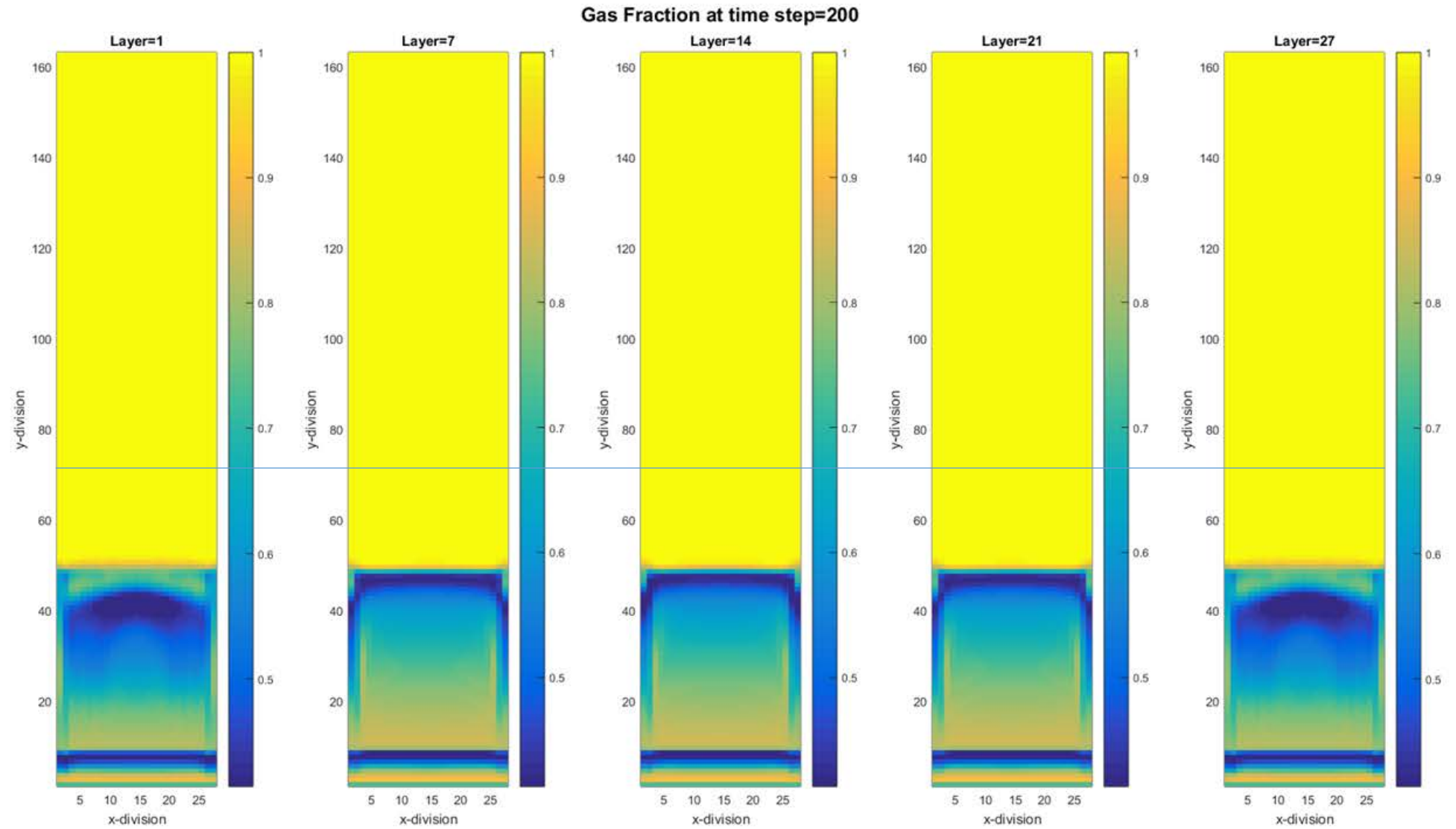
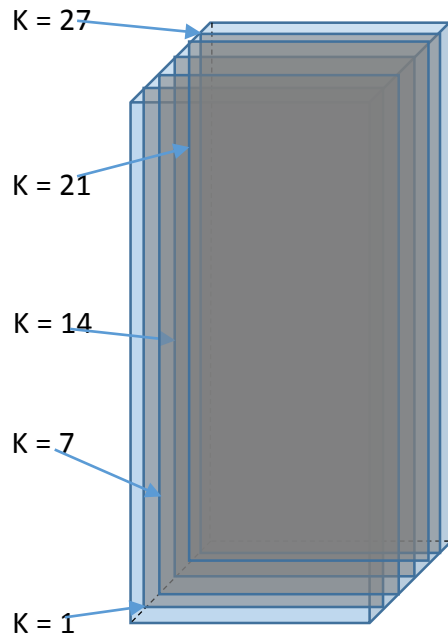


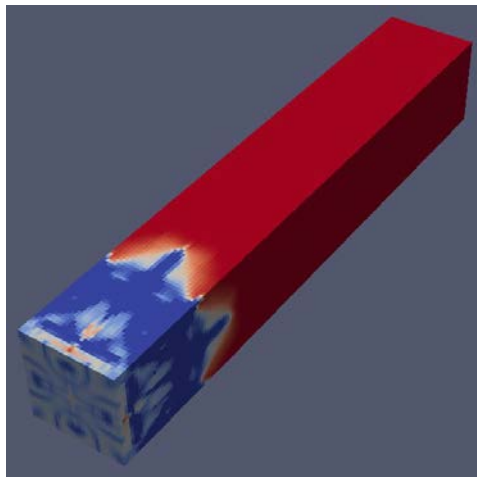
Late-time
Time step 4000

Five cross-sections used to show details in 2-D

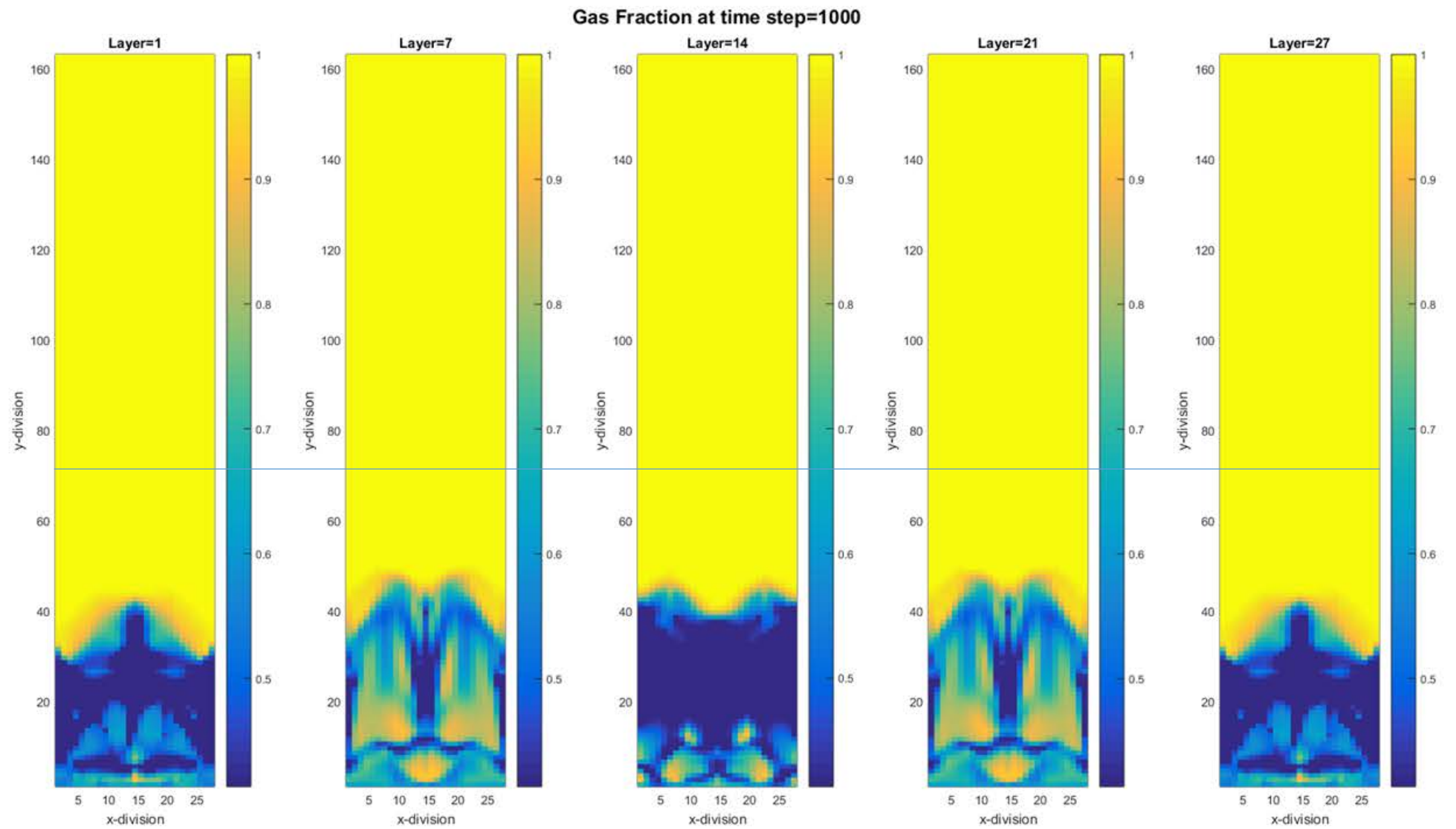
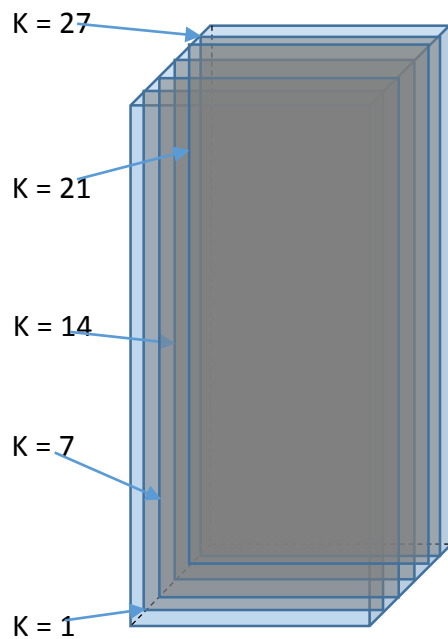


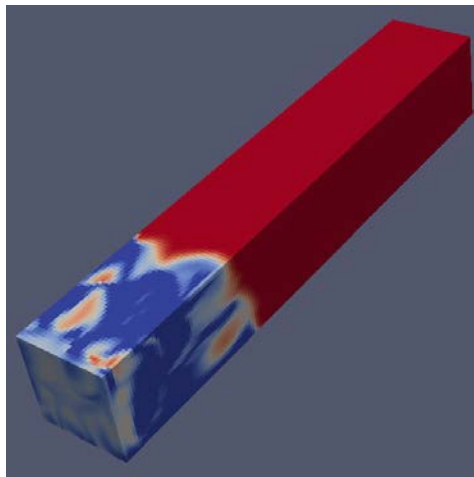
Early-time
Time step 200



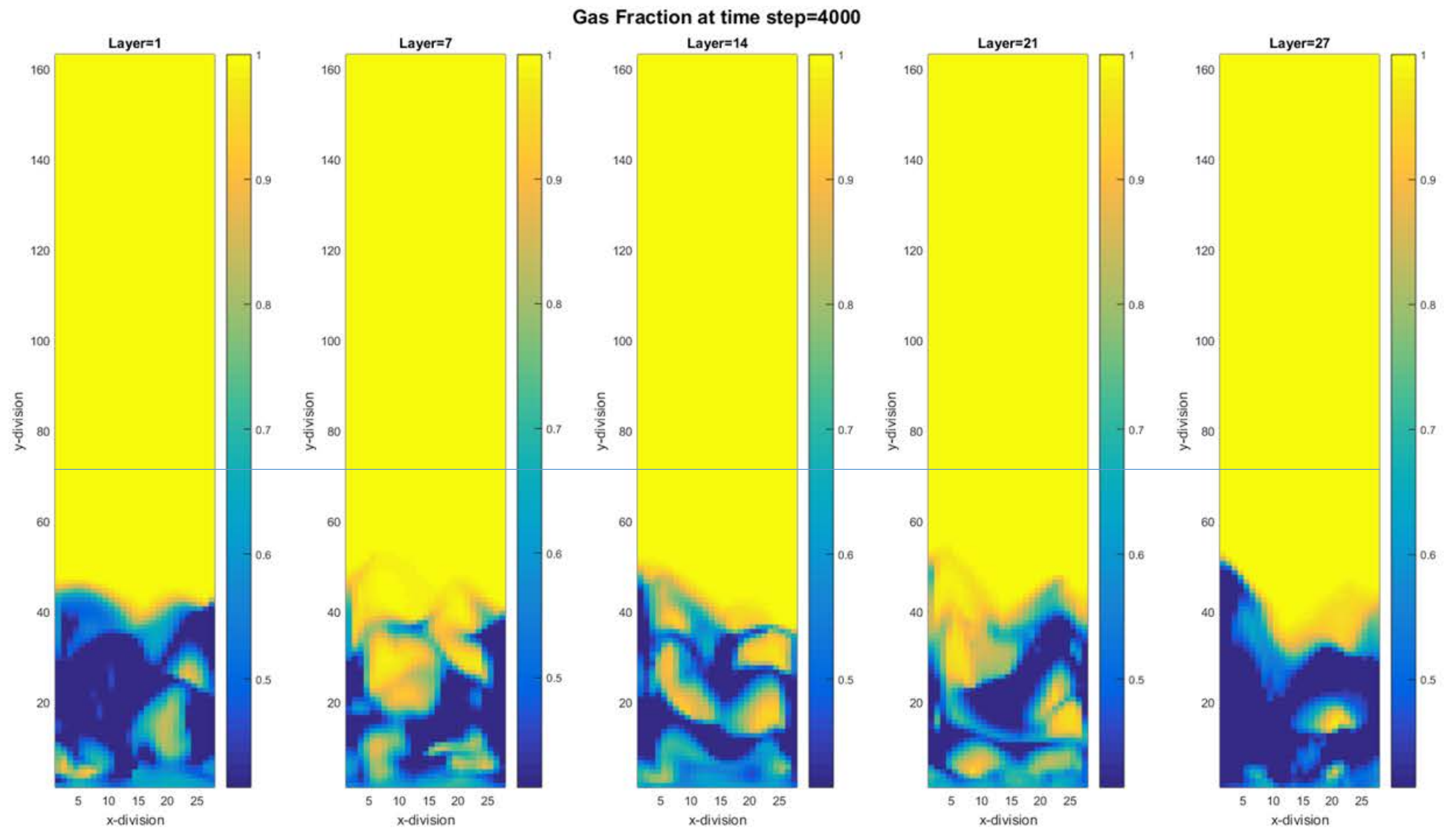
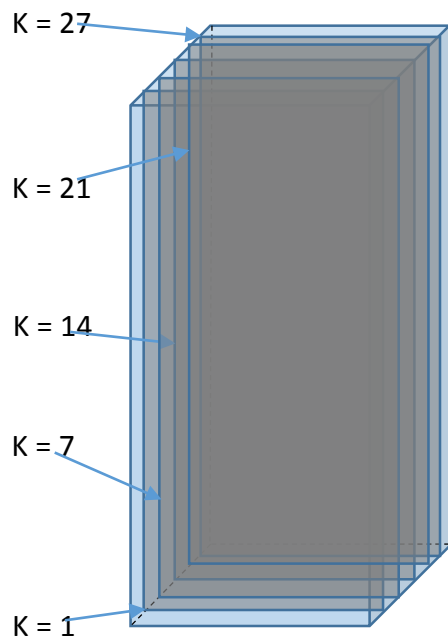


Mid-time
Time step 1000



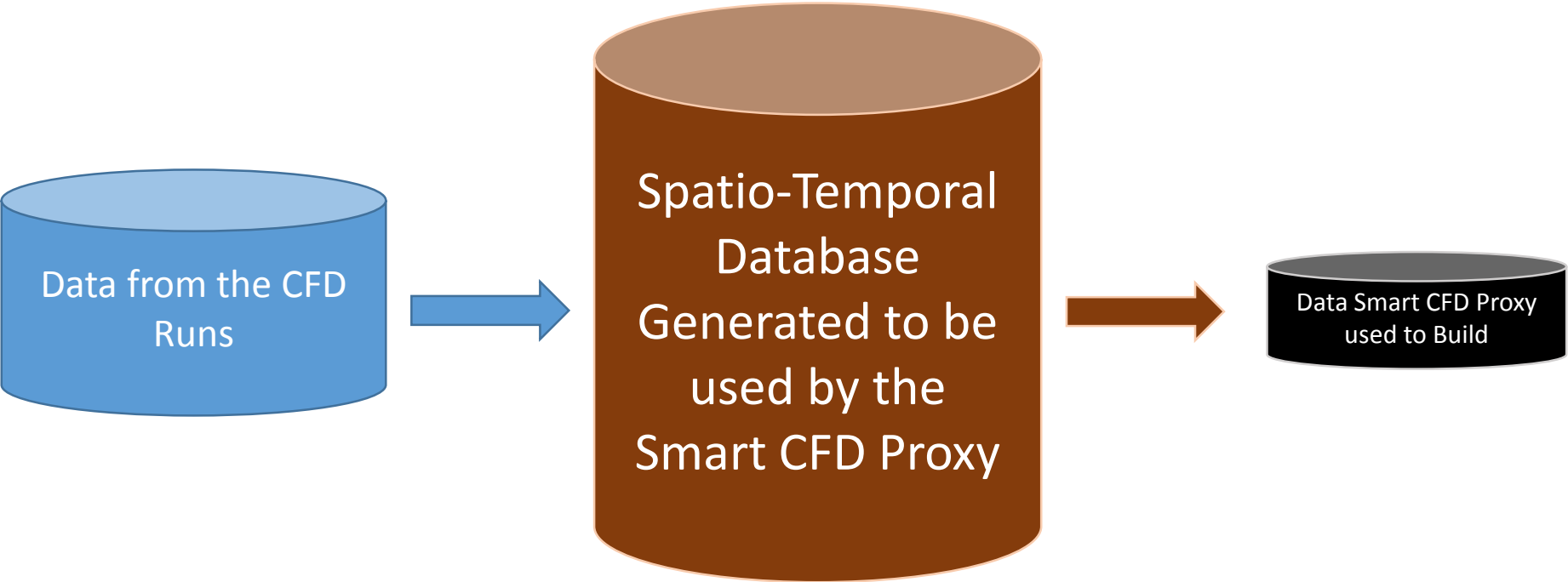


**Late-time
Time step 4000**



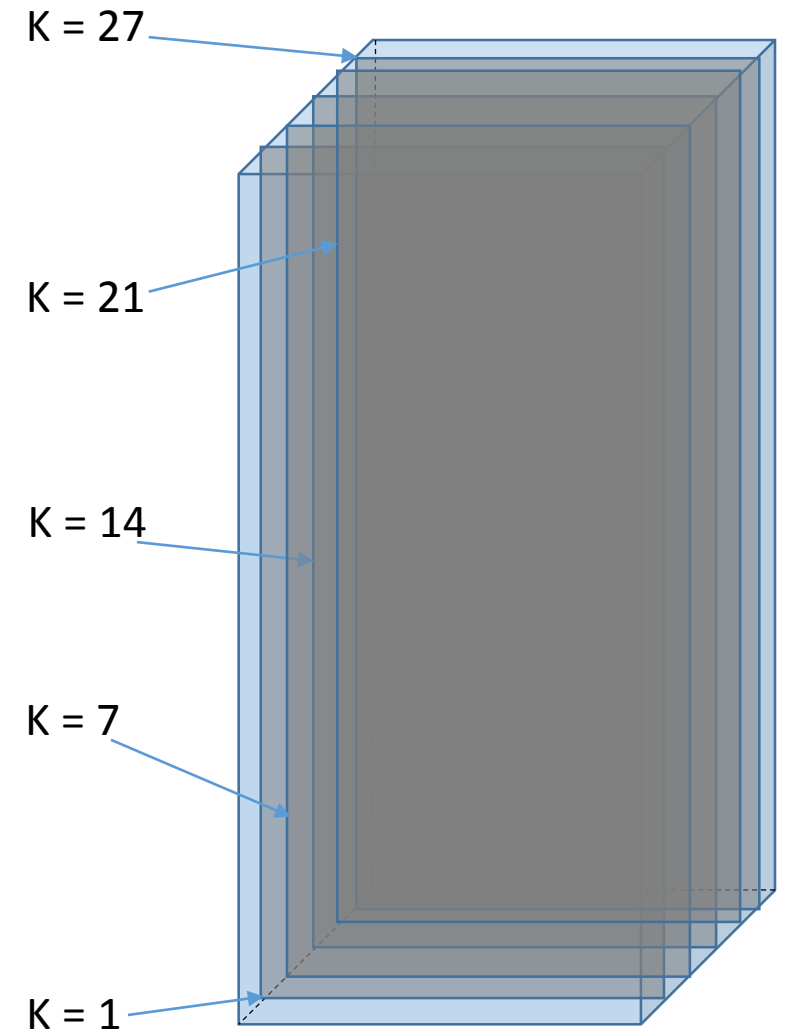
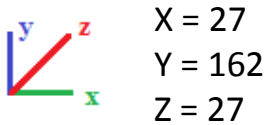
Data from the CFD Runs

Key factors of multi-phase gas-solid system	
Volume fraction (ϵ)	Gas Density (ρ)
Velocity vector of gas (u, v, w)	Particle diameter (d)
Velocity vector of solid (u, v, w)	Maximum packing volume fraction (ϵ^*)
Pressure field of gas (P)	Location to the boundaries
Pressure field of solid(P)	Location to the interface(x, y, z)
Time (t)	Viscosity (μ and λ)
	Gravity force (g)

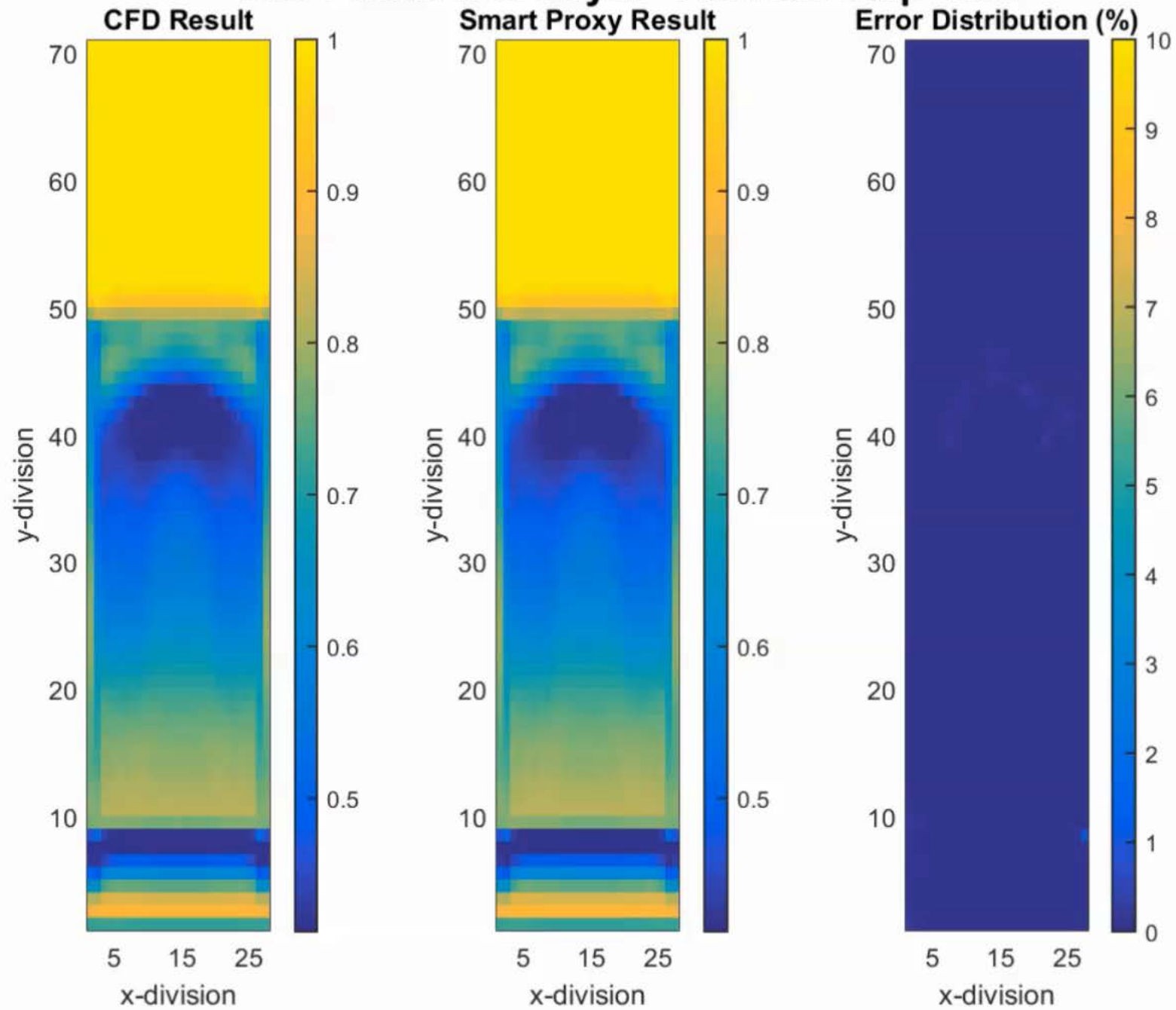


Displaying Smart CFD Proxy

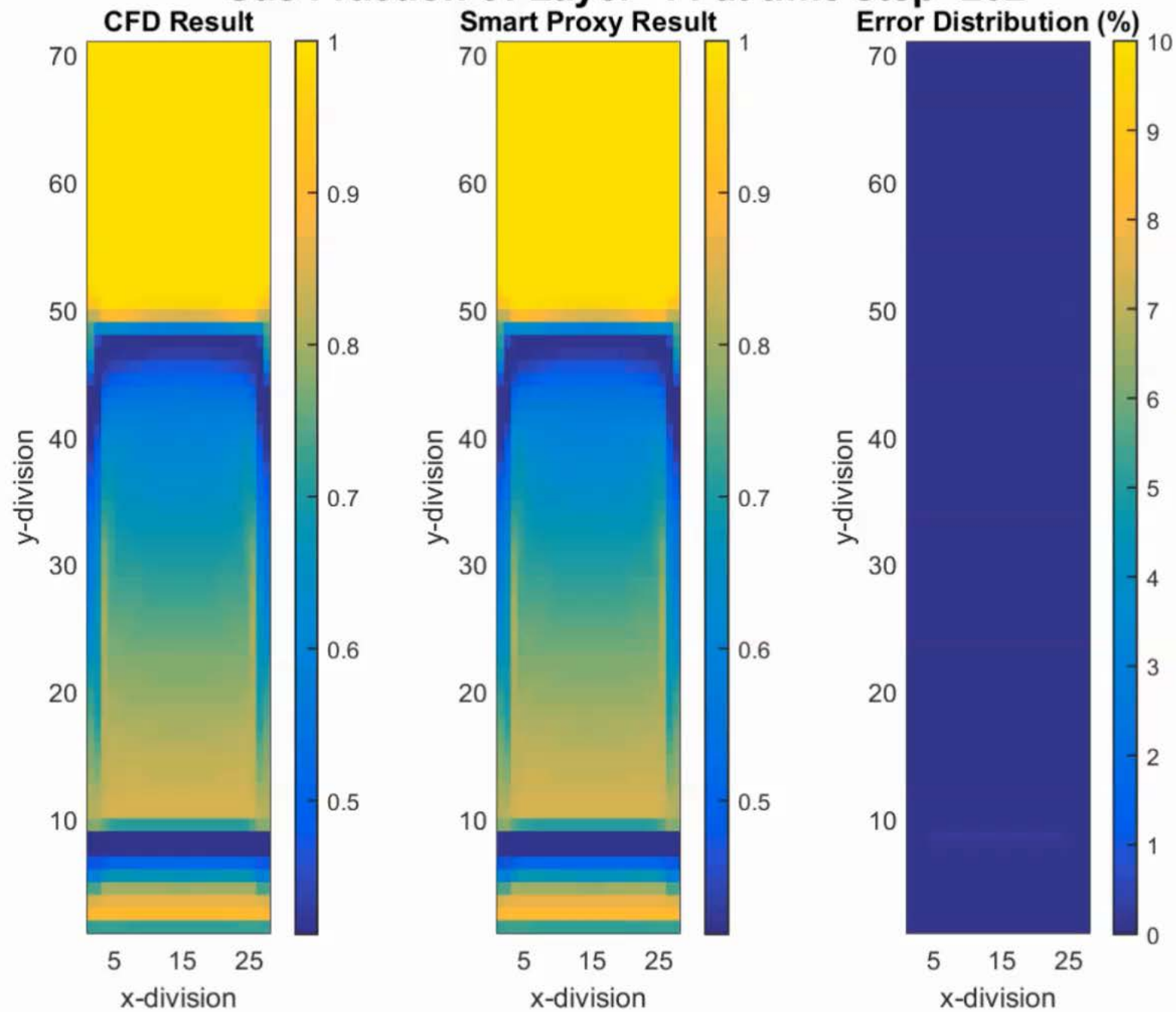
- Results are displayed in 5 different cross-sections:
- $K = 1, 7, 14, 21, 27$



Gas Fraction of Layer=1 at time step=202



Gas Fraction of Layer=14 at time step=202



Thank you

Extra Slides

Part One – Non-Cascading Results

- In Non-Cascading scheme at every time-step actual CFD results are used as input and the trained Smart Proxy forecasts the results for the next time-step.
- No error accumulation takes place in this scheme.
- This scheme, as the first step, explores the learning capabilities of the Smart Proxy.

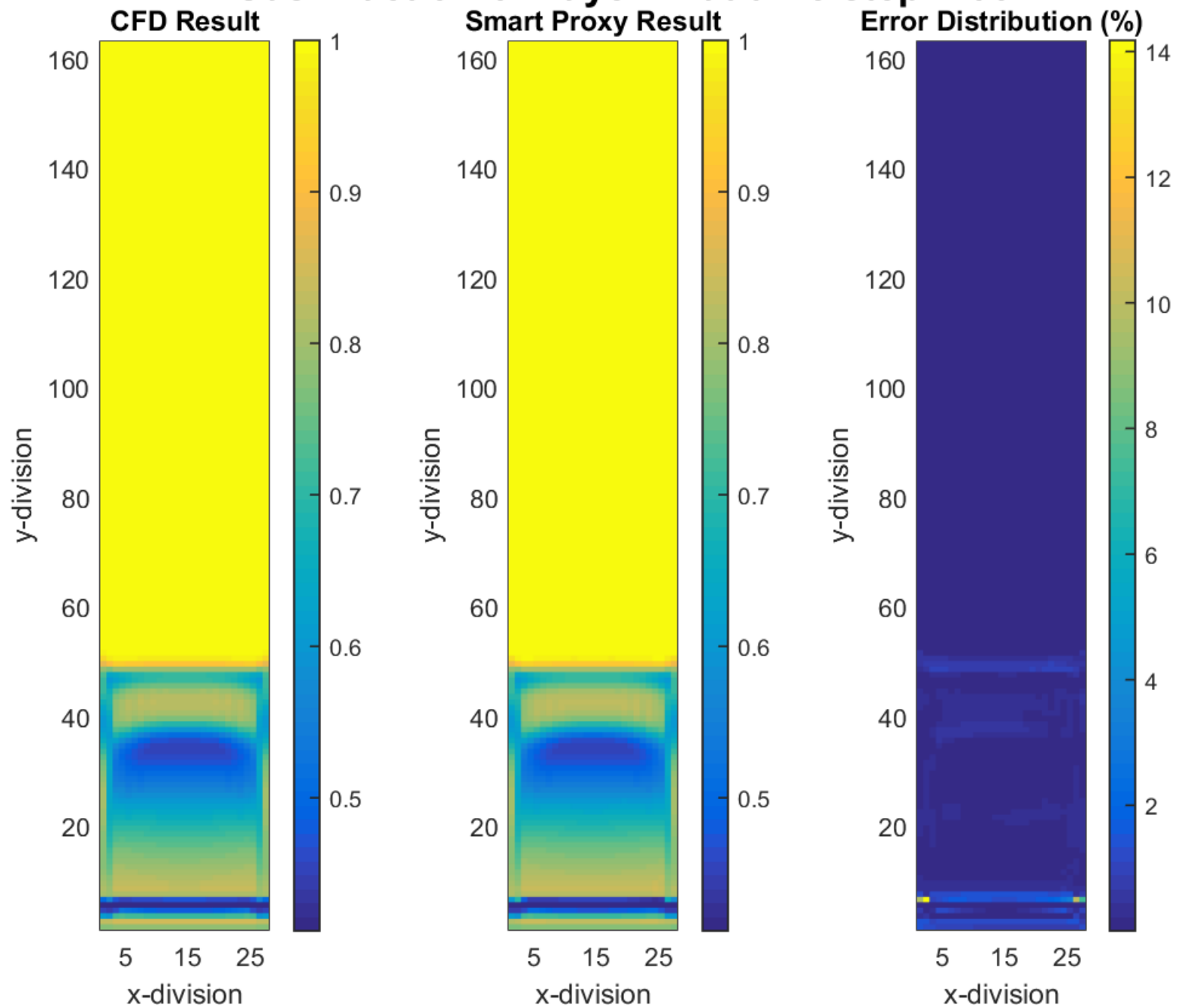
Early Time Results

- Smart Proxy is generated for the early time in the system when the bubble behavior is reasonably calm and non-chaotic.

Results of Layer 1 – Gas Fraction

- Model was trained using Time-step 101 as input and Time-step 102 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 102 through Time-step 126
- Results generated by the Smart Proxy are compared with actual CFD output.

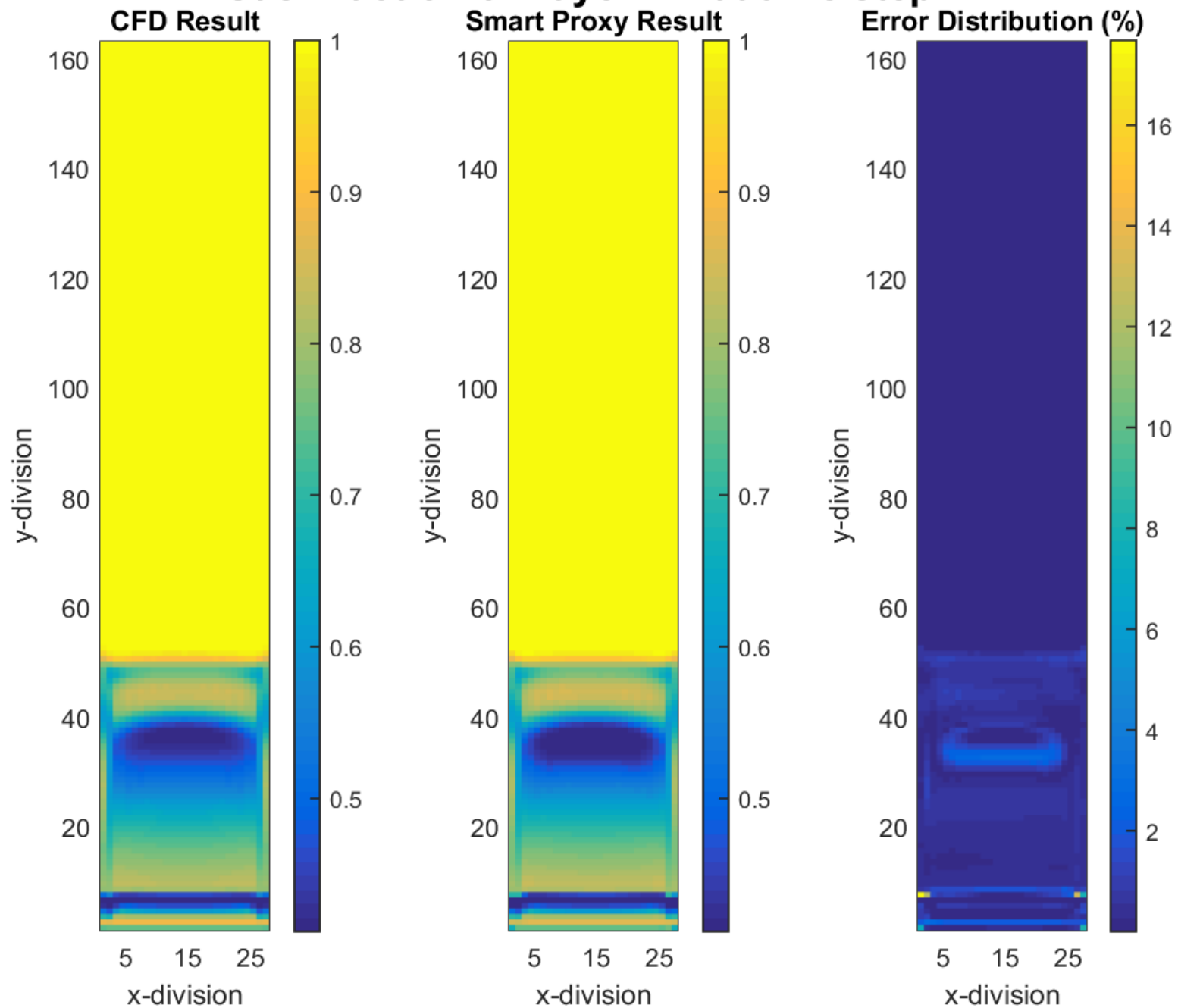
Gas Fraction of Layer=1 at time step=106



Results of Layer 27 – Gas Fraction

- Model was trained using Time-step 101 as input and Time-step 102 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 102 through Time-step 126
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=27 at time step=112



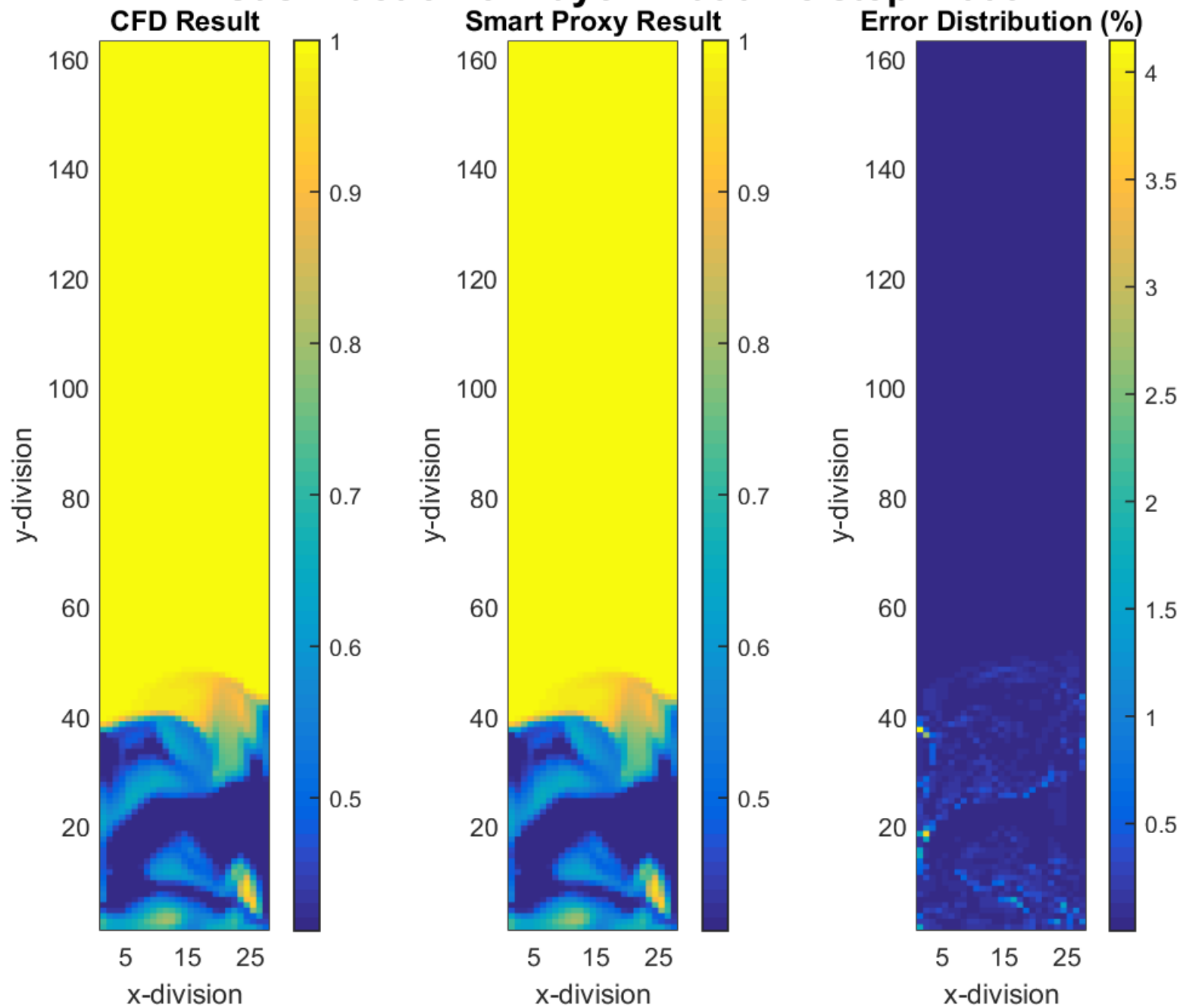
Multiple bubbles in the system

- The process is now repeated when several bubbles have already been formed in the system.

Results of Layer 1 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4001 through Time-step 4040
- Results generated by the Smart Proxy are compared with actual CFD output.

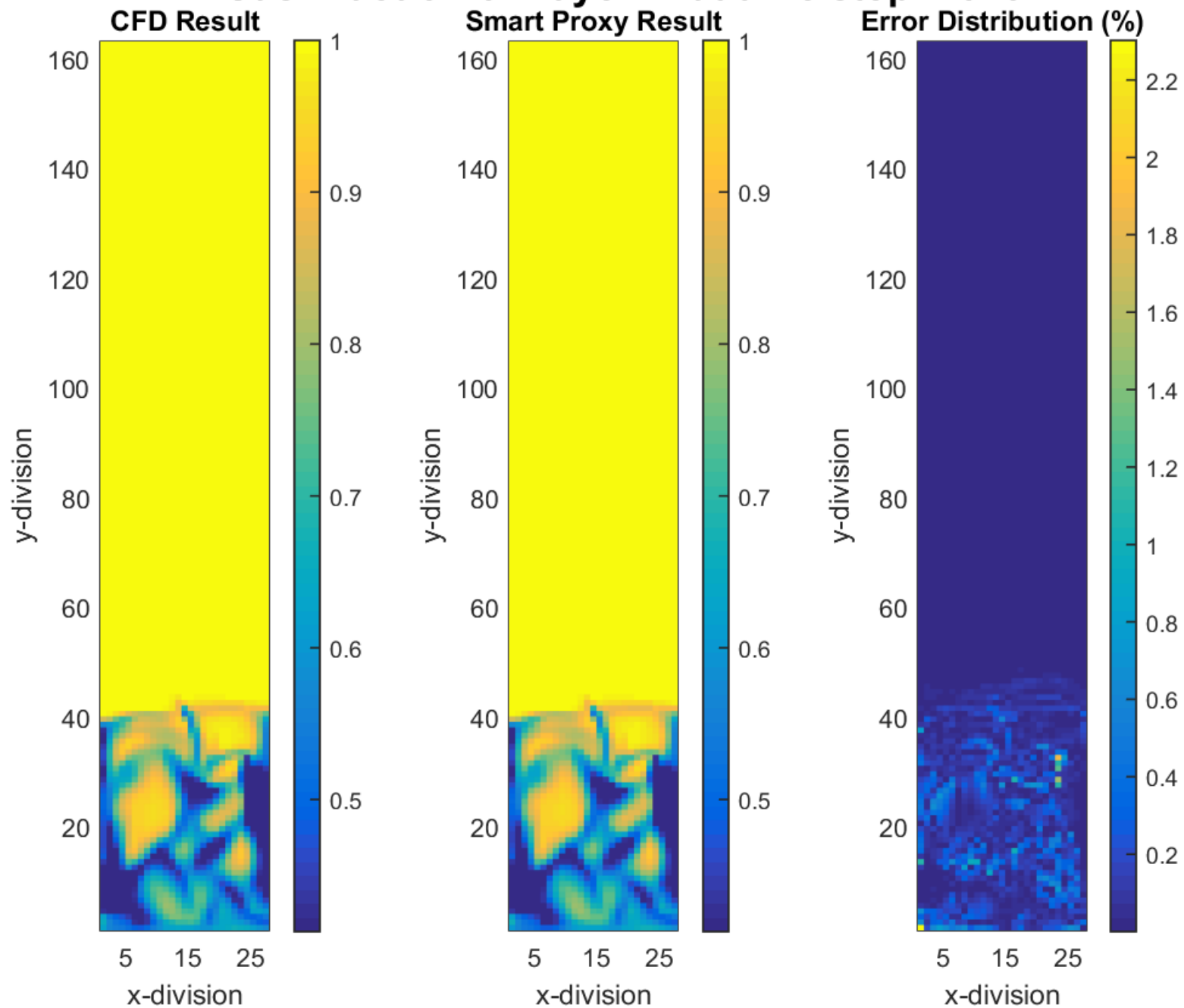
Gas Fraction of Layer=1 at time step=4006



Results of Layer 7 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4001 through Time-step 4040
- Results generated by the Smart Proxy are compared with actual CFD output.

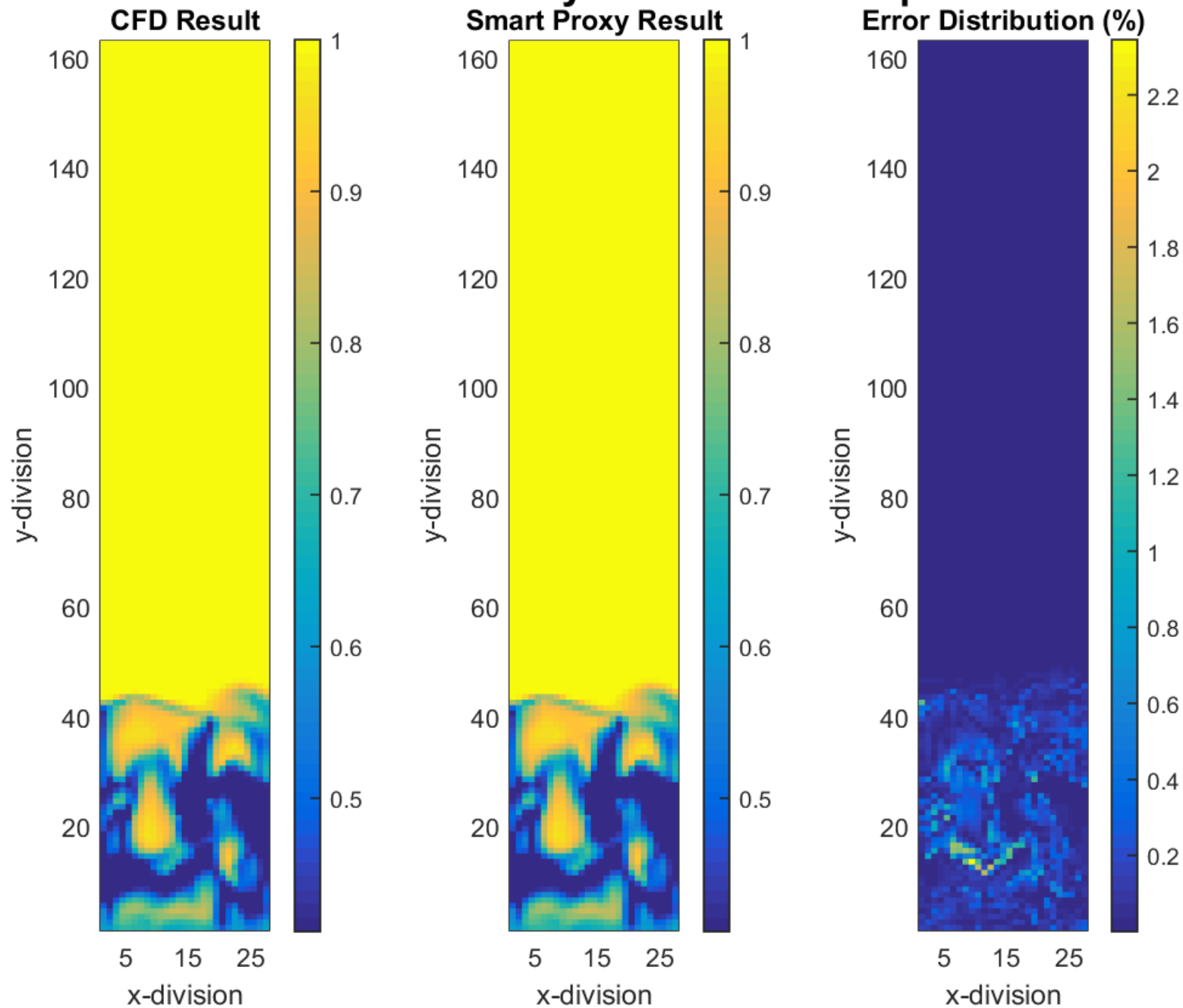
Gas Fraction of Layer=7 at time step=4016



Results of Layer 14 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4001 through Time-step 4040
- Results generated by the Smart Proxy are compared with actual CFD output.

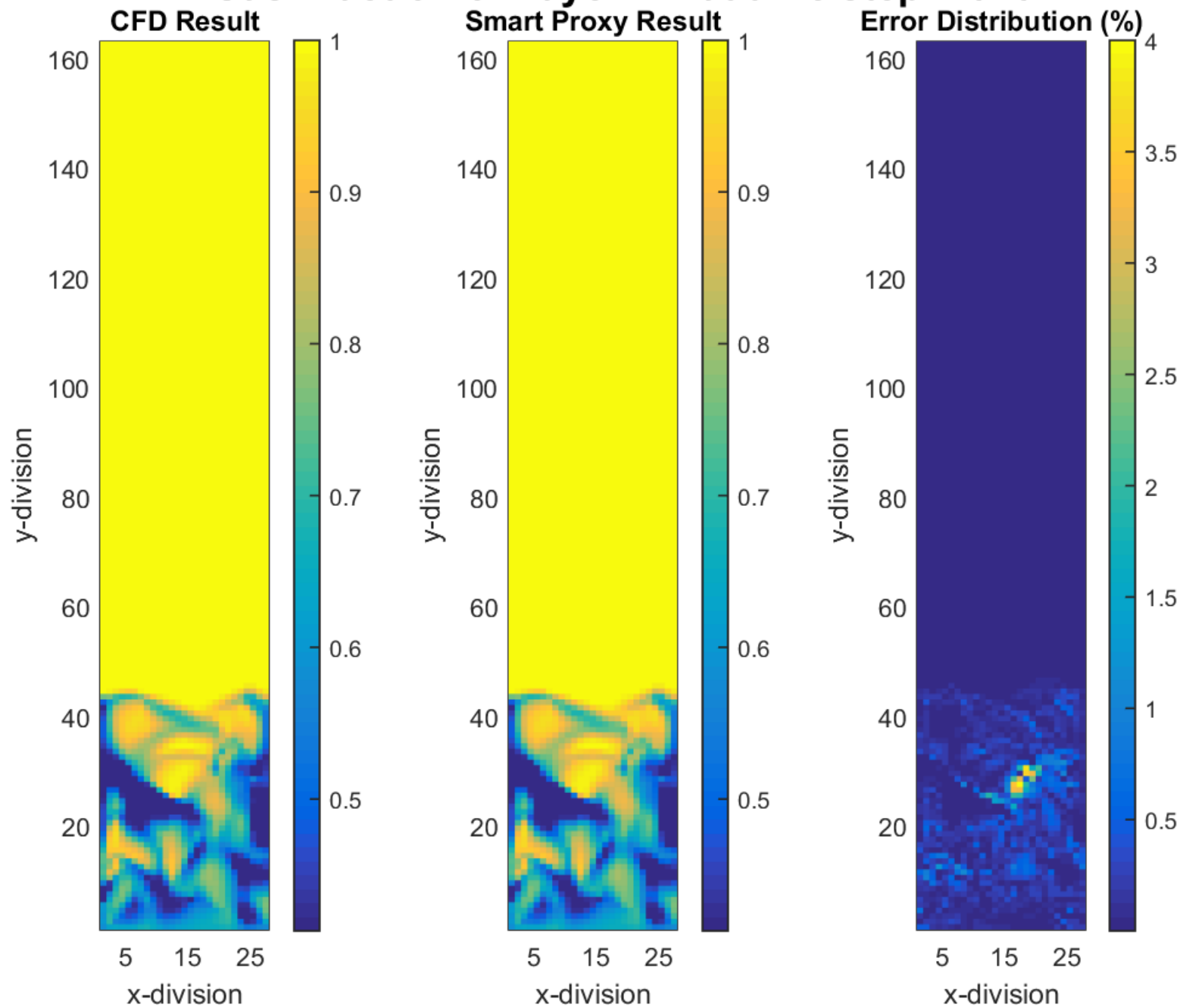
Gas Fraction of Layer=14 at time step=4014



Results of Layer 21 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4001 through Time-step 4040
- Results generated by the Smart Proxy are compared with actual CFD output.

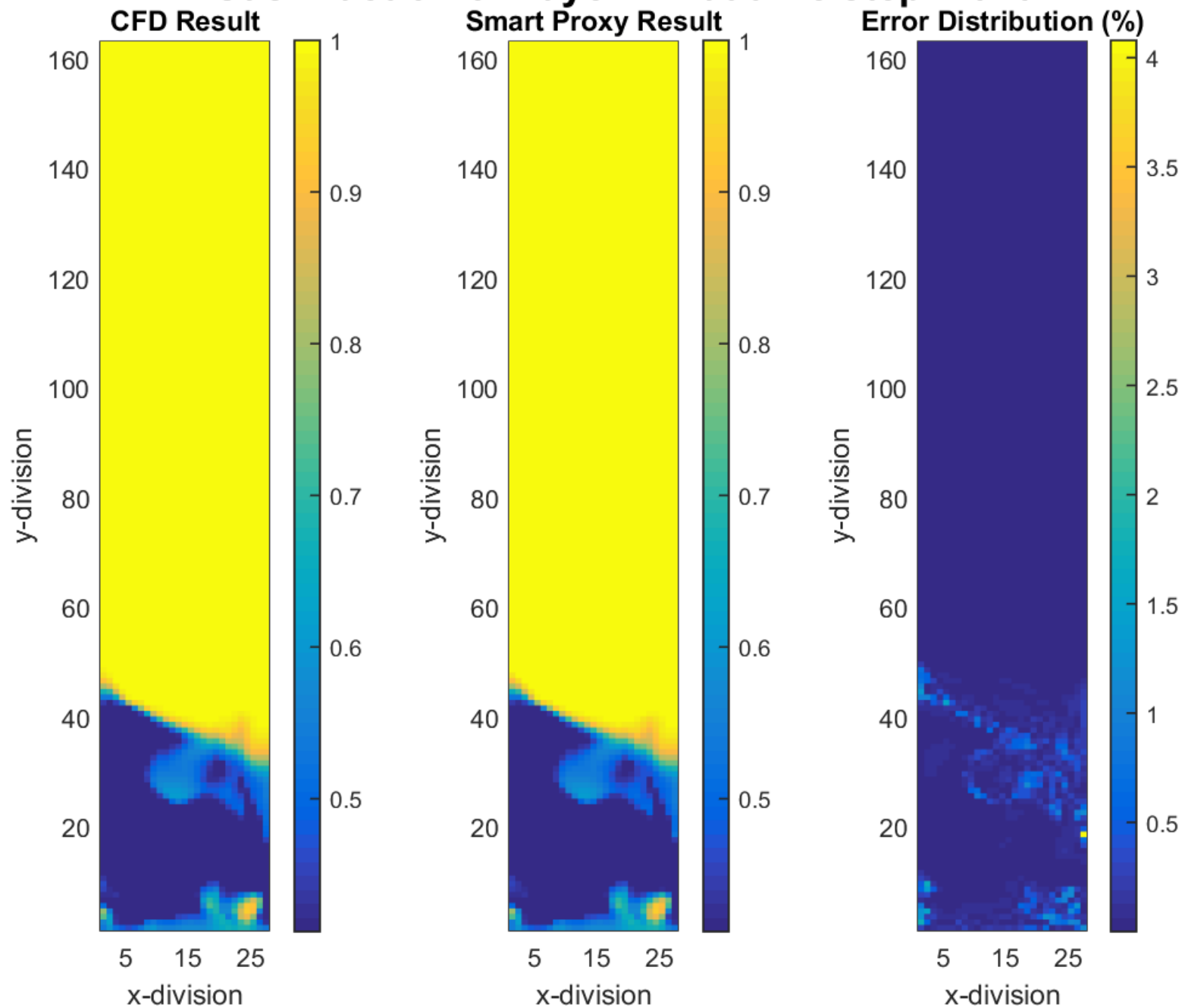
Gas Fraction of Layer=21 at time step=4020



Results of Layer 27 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4001 through Time-step 4040
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=27 at time step=4020



Part Two – Cascading Results

- In Cascading scheme at every time-step results from the Smart Proxy are used as input and the trained Smart Proxy forecasts the results for the next time-step. (Smart Proxy receives feedback from itself)
 - **Models are developed for all output variables in CFD such as Void Fraction, and gas/solid Pressure and Velocity.**
- In this scheme the error has the opportunity to accumulate.
- Limits of the information content of each time-step is explored in this scheme.

Early Time Results

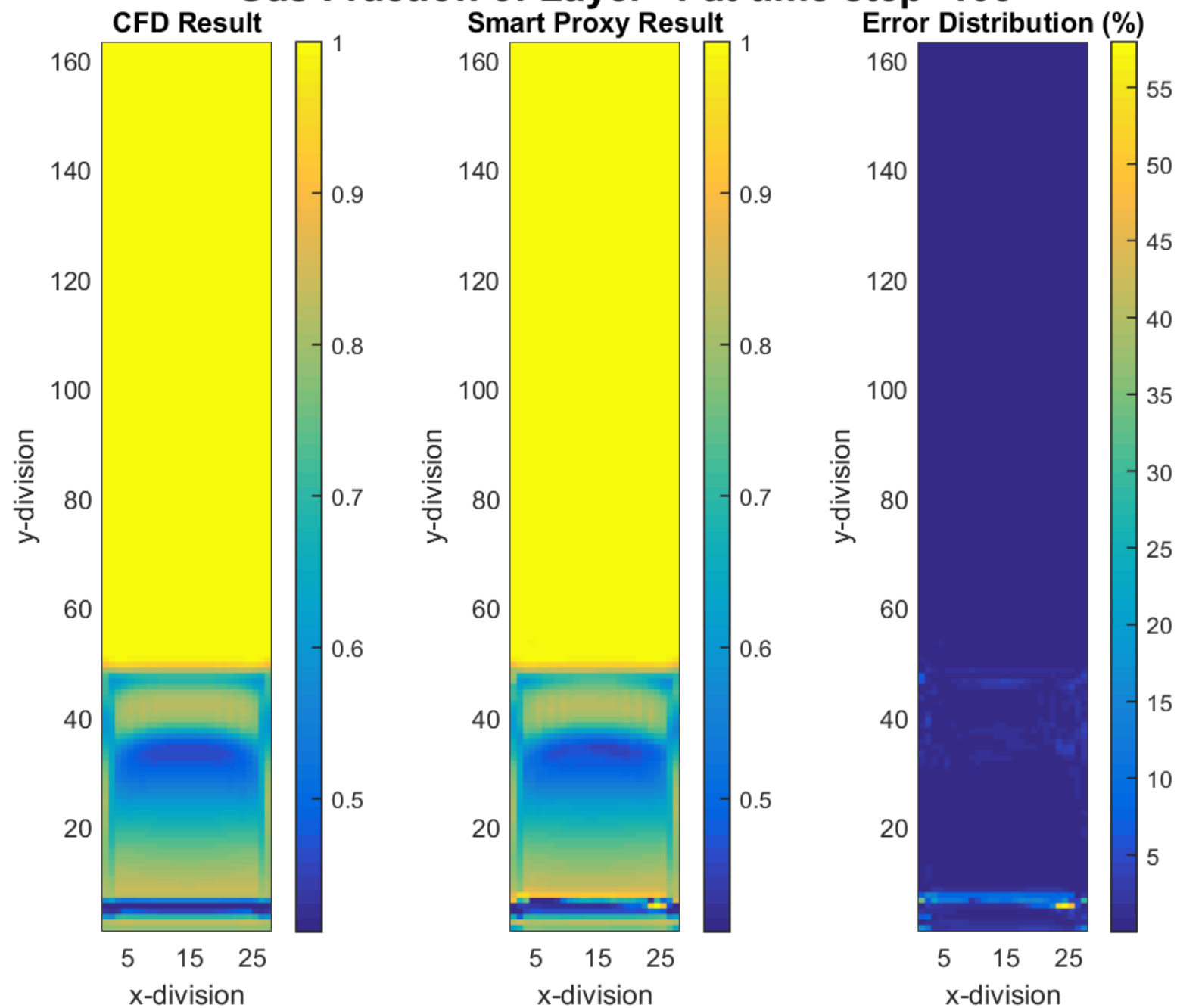
- Smart Proxy is generated for the early time in the system when the bubble behavior is reasonably calm and non-chaotic.

Results of Layer 1 – Gas Fraction

- Model was trained using Time-step 101 as input and Time-step 102 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 101 through Time-step 105
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=1 at time step=105

C

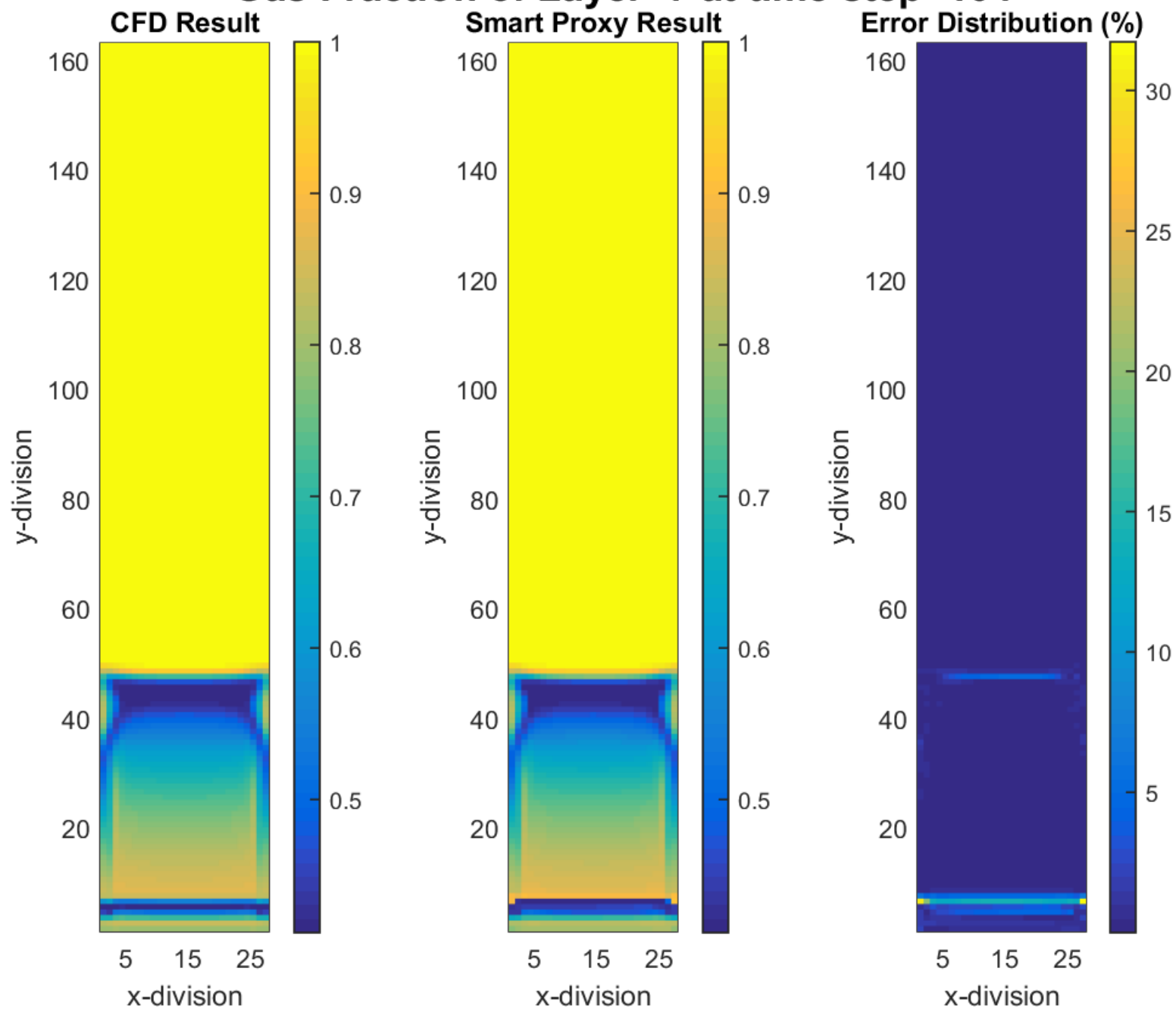


Results of Layer 7 – Gas Fraction

- Model was trained using Time-step 101 as input and Time-step 102 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 101 through Time-step 105
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=7 at time step=104

C

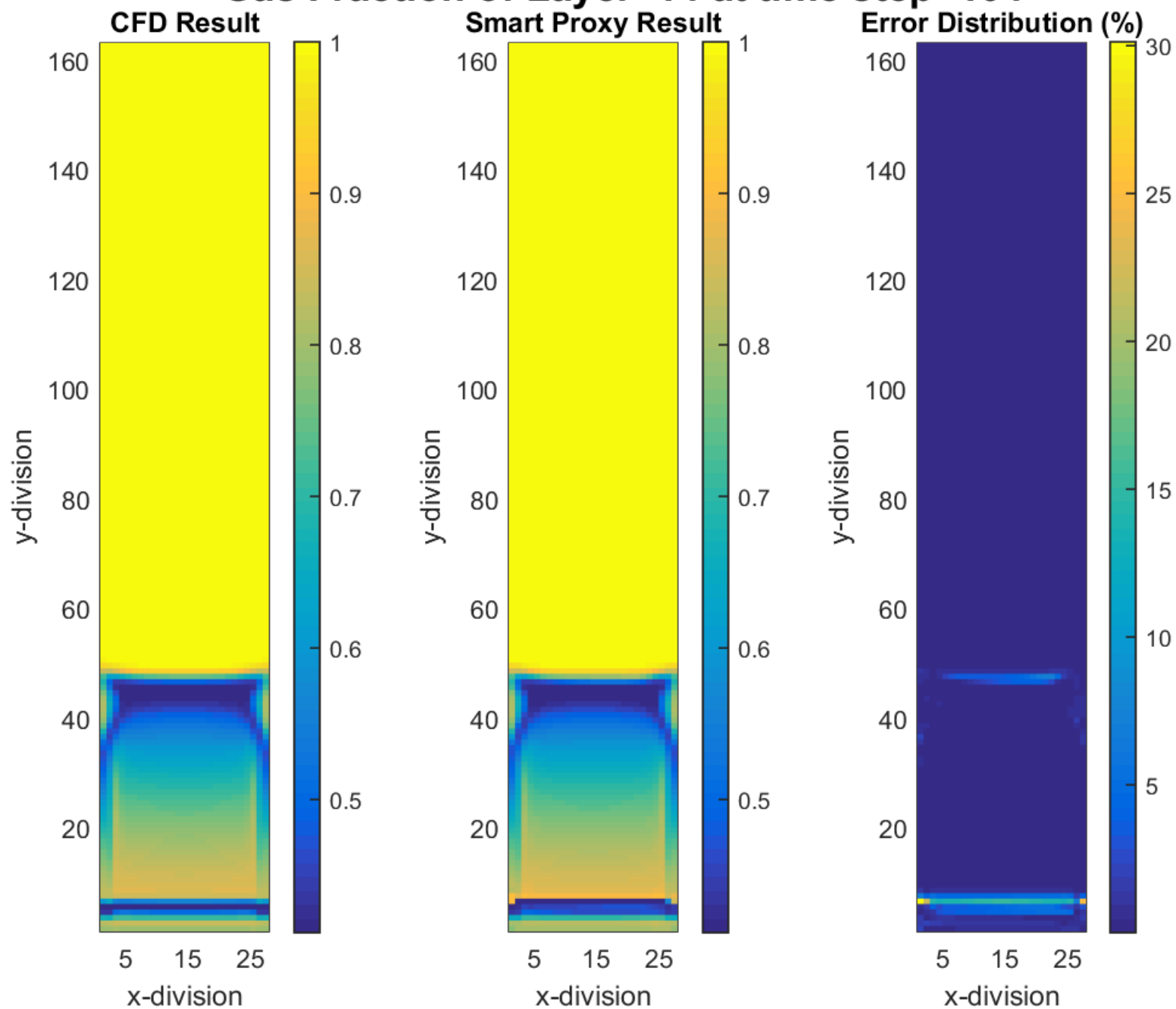


Results of Layer 14 – Gas Fraction

- Model was trained using Time-step 101 as input and Time-step 102 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 101 through Time-step 105
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=14 at time step=104

C

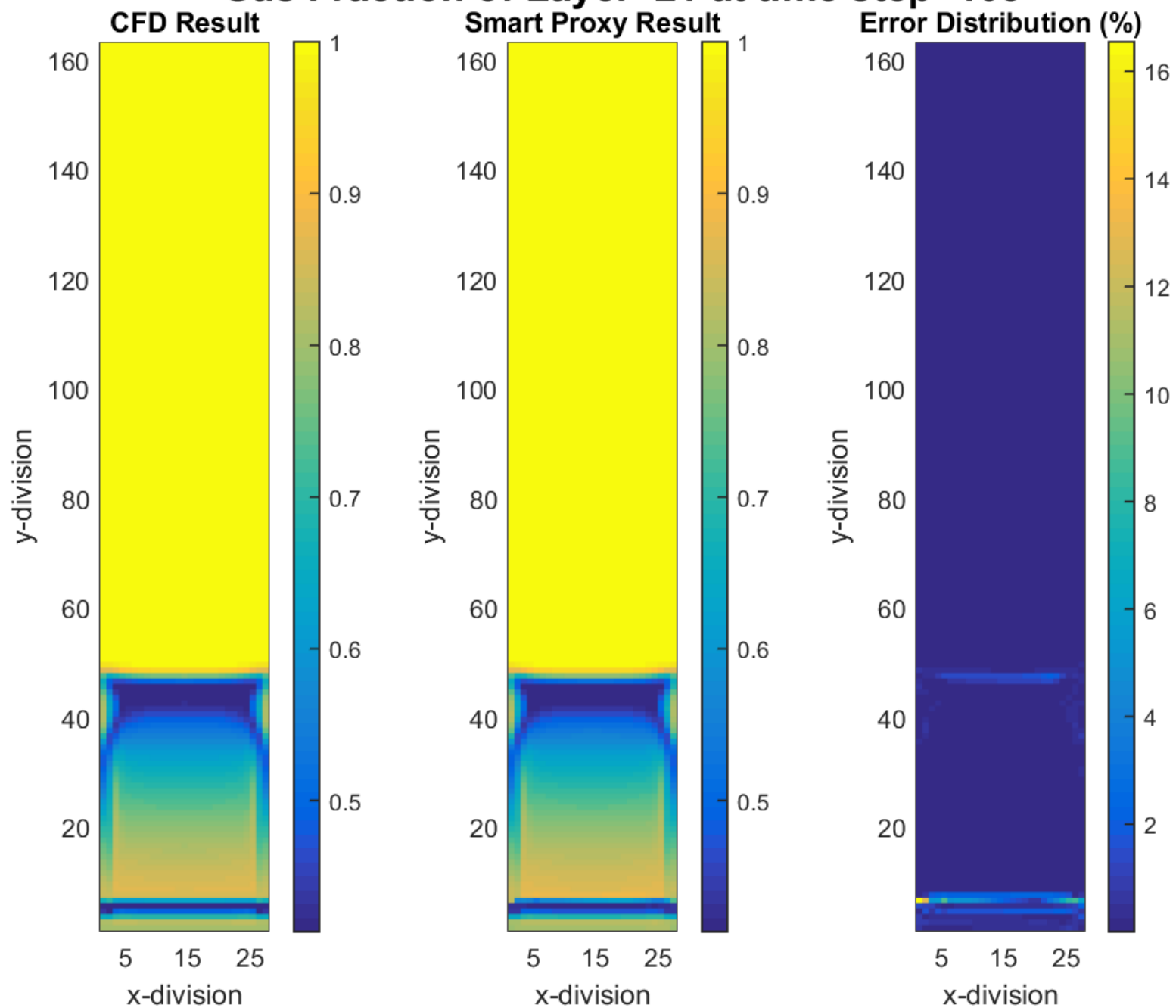


Results of Layer 21 – Gas Fraction

- Model was trained using Time-step 101 as input and Time-step 102 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 102 through Time-step 126
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=21 at time step=103

C

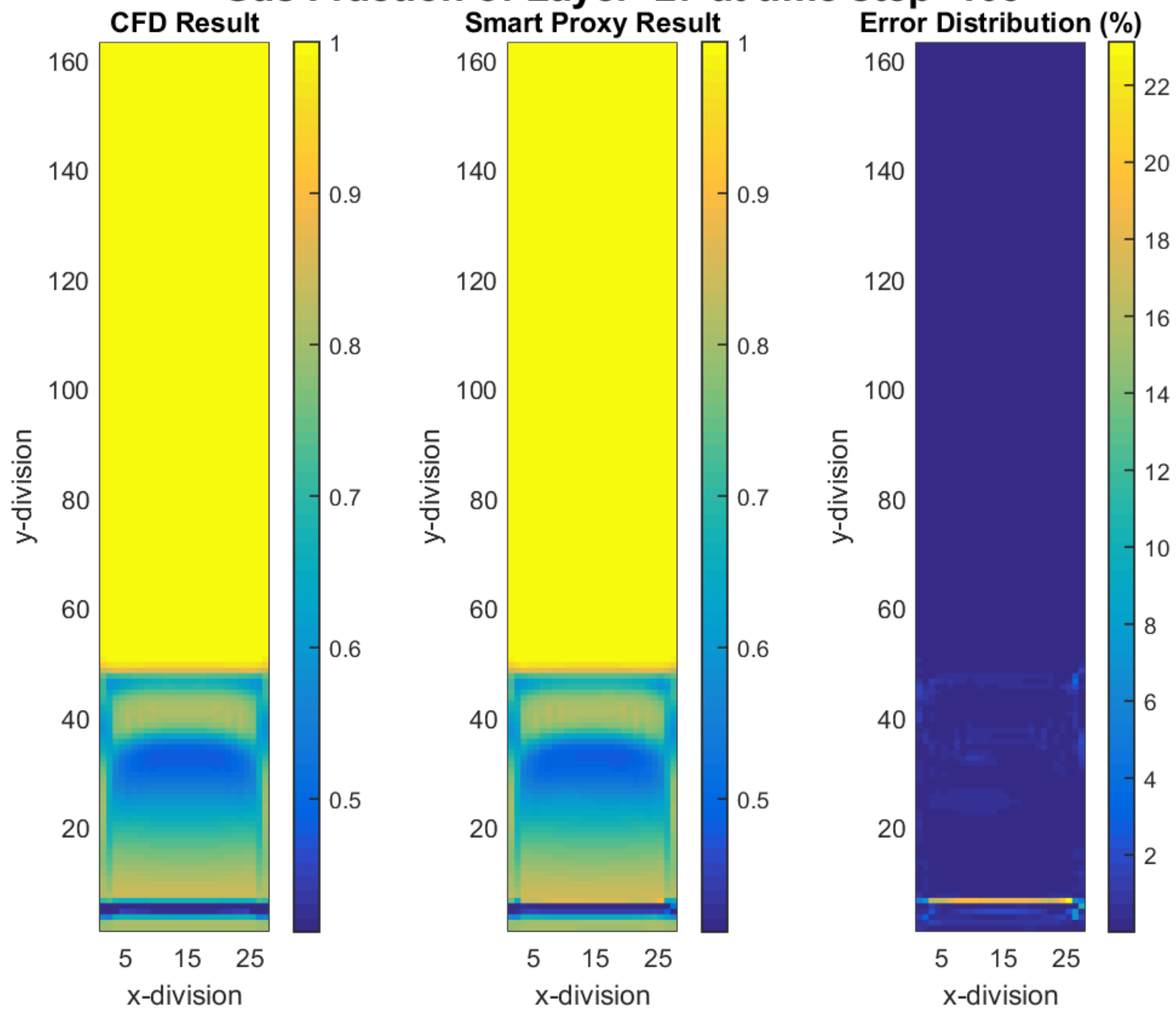


Results of Layer 27 – Gas Fraction

- Model was trained using Time-step 101 as input and Time-step 102 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 101 through Time-step 105
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=27 at time step=103

C



Multiple bubbles in the system

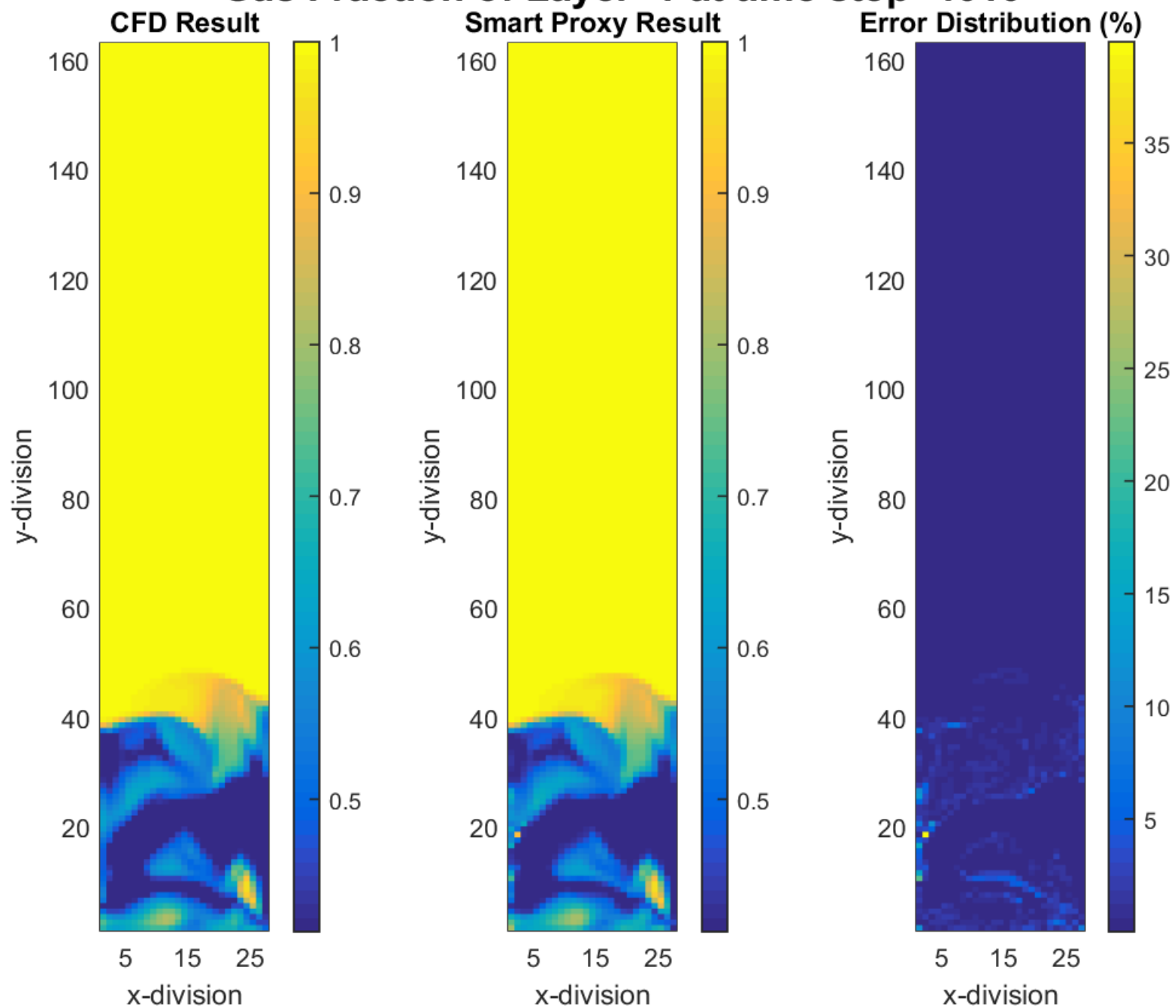
- The process is now repeated when several bubbles have already been formed in the system.

Results of Layer 1 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4002 through Time-step 4014
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=1 at time step=4010

C

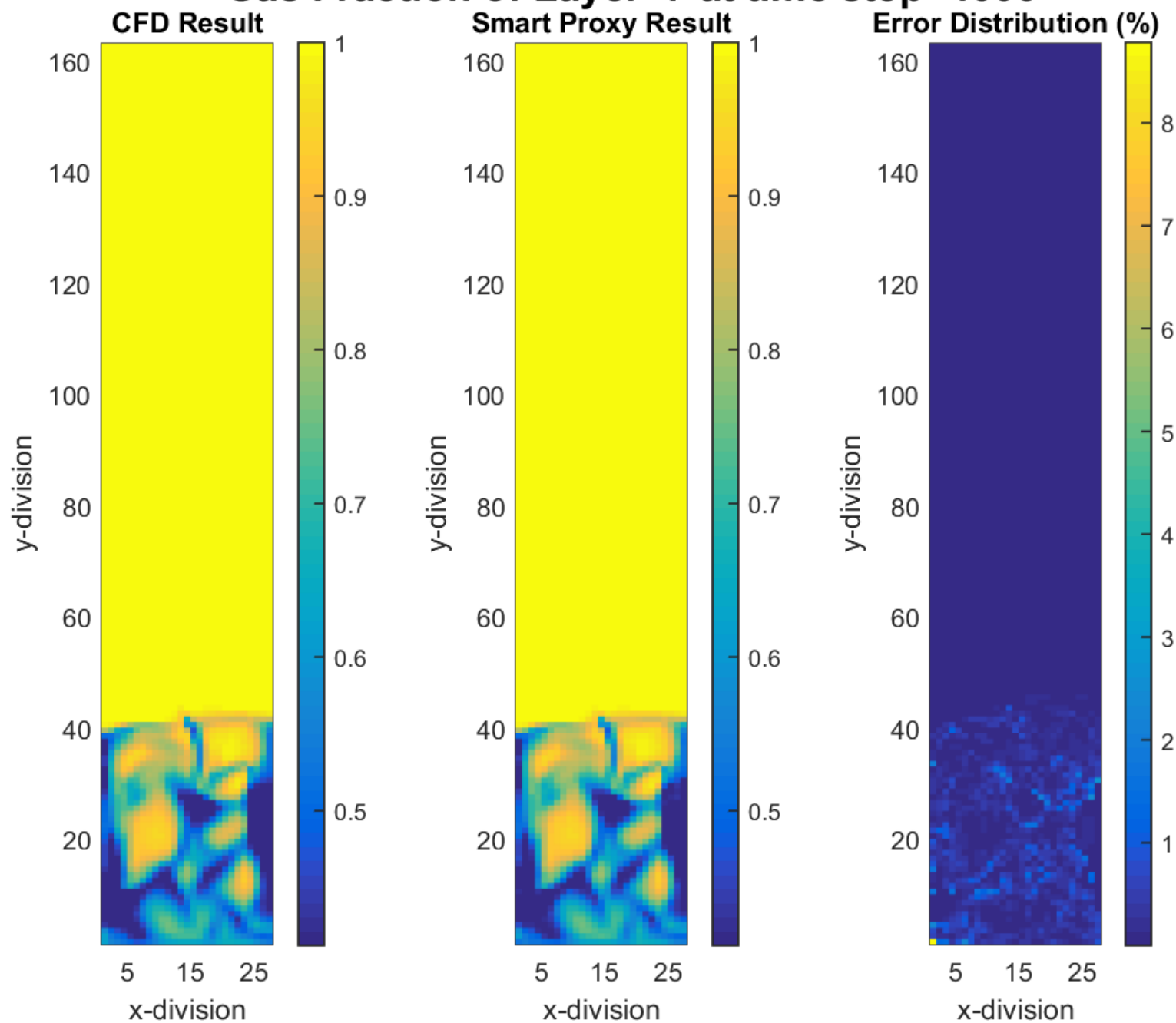


Results of Layer 7 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4002 through Time-step 4010
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=7 at time step=4006

C

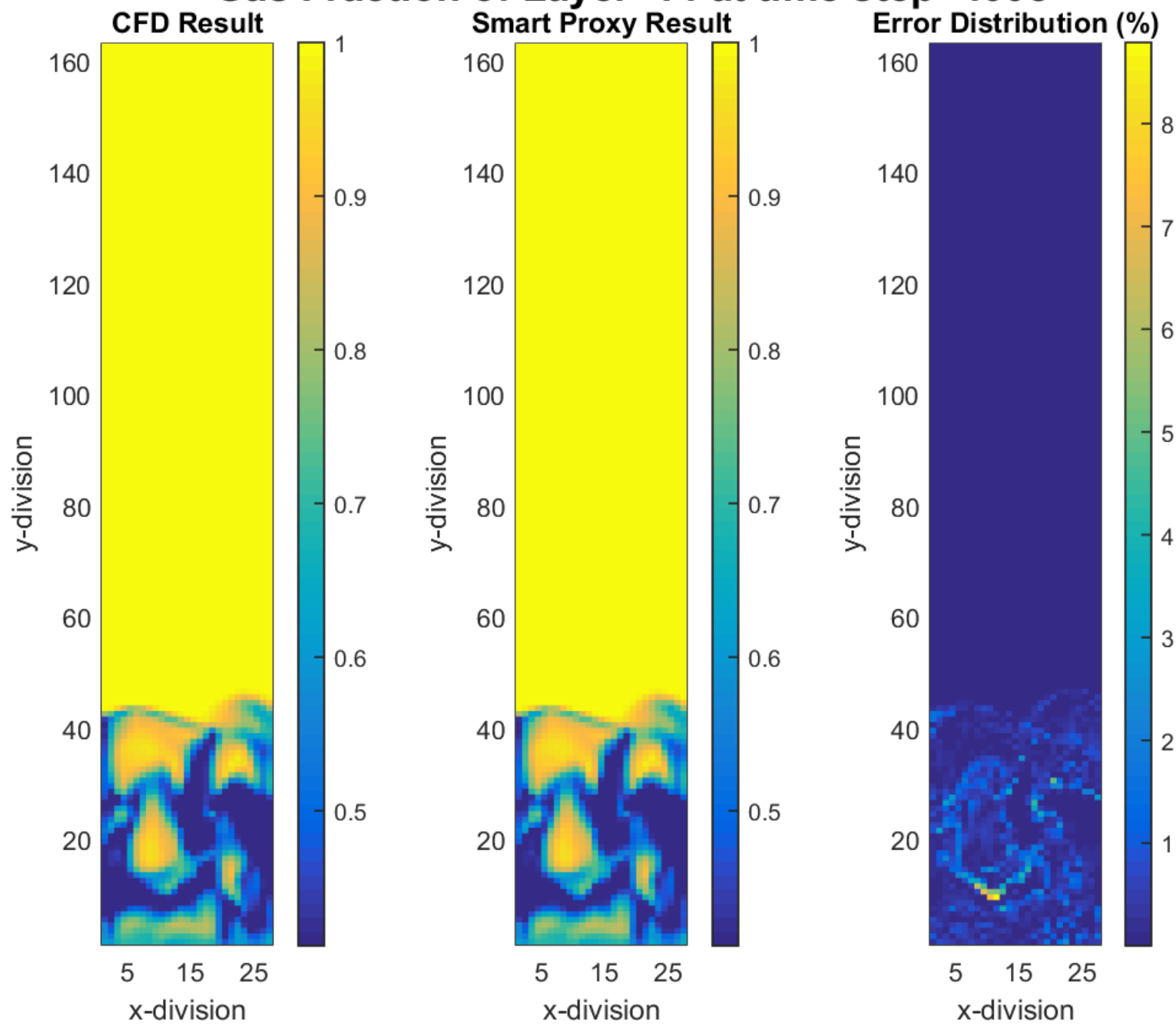


Results of Layer 14 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4002 through Time-step 4024
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=14 at time step=4008

C

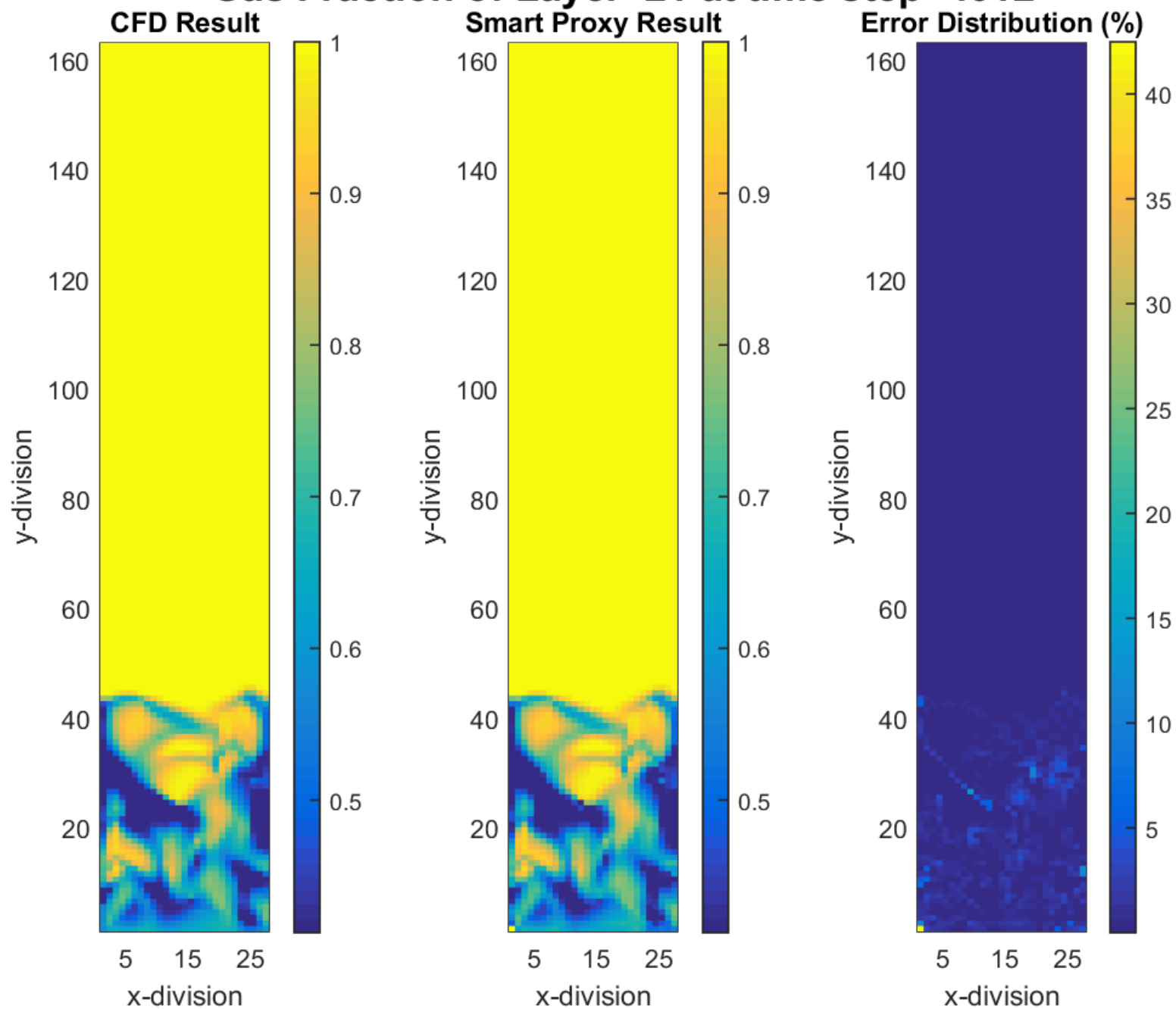


Results of Layer 21 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4002 through Time-step 4020
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=21 at time step=4012

C

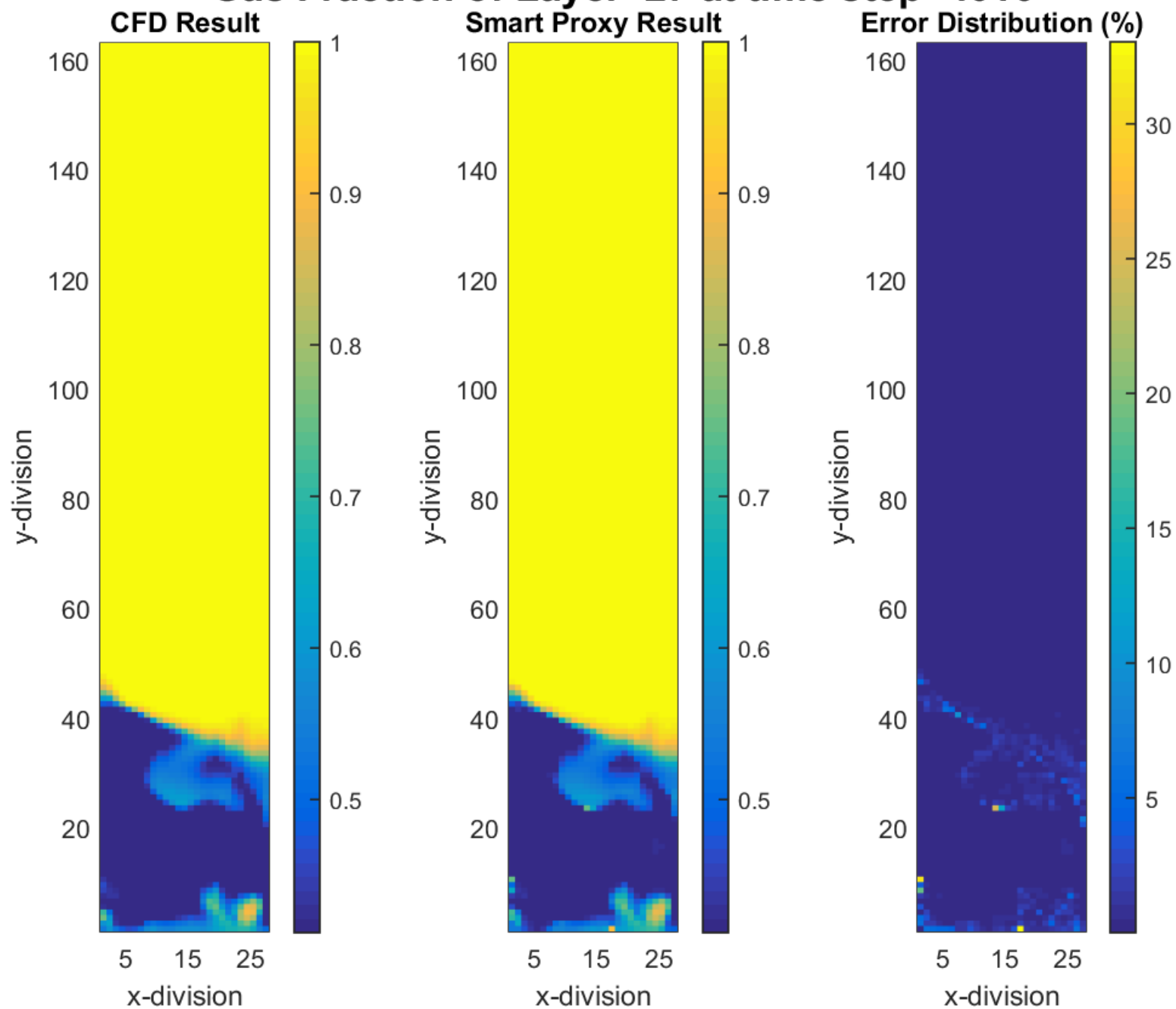


Results of Layer 27 – Gas Fraction

- Model was trained using Time-step 4001 as input and Time-step 4002 as output.
 - Trained on 75% of data and validated using the remaining 25%
- Trained model was deployed in forecast mode and generated Time-step 4002 through Time-step 4012
- Results generated by the Smart Proxy are compared with actual CFD output.

Gas Fraction of Layer=27 at time step=4010

C



Paradigm Shift in Proxy Modeling

- **VERY LARGE AMOUNTS OF DATA** is carefully extracted from a handful of runs of the original model.
- The process of data extraction from several runs of the original model is **ENGINEERED** such that it can be used to train, calibrate and validate a smart, new model.

Paradigm Shift in Proxy Modeling

- The Smart Proxy model is developed using **MACHINE LEARNING** that is now the main paradigm for treating and handling **“BIG DATA”**.
- The Smart Proxy learns to mimic the behavior of the original model with all its complexities and intricacies.
- Once developed, the Smart Proxy Model runs at very low computational cost (thousands of simulation runs in minutes).

The Spatio-Temporal Database

Tier One Neighbor Cells

<i>Cell_{ijk} at time step n</i>								
ε_g	P	P_s	u_g	v_g	w_g	u_s	v_s	w_s

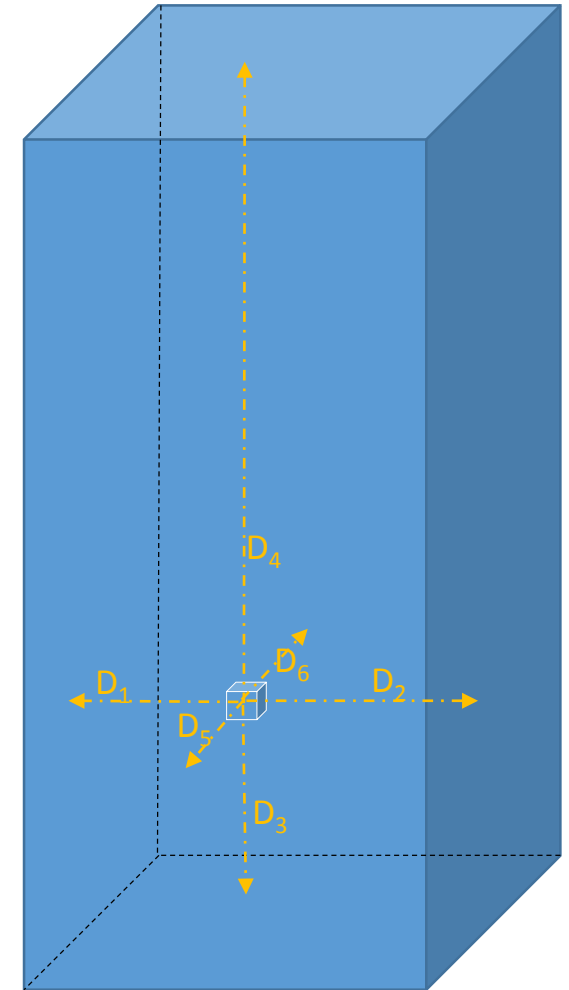
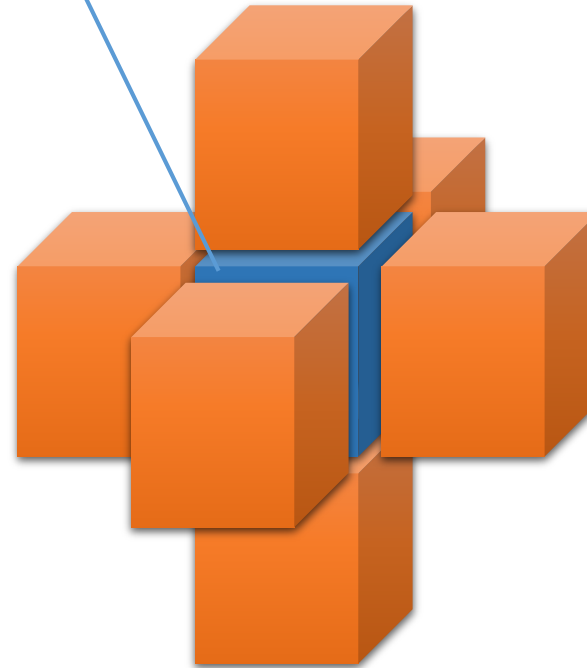
<i>Tier Cell_{ijk} at time step n</i>								
ε_g	P	P_s	u_g	v_g	w_g	u_s	v_s	w_s

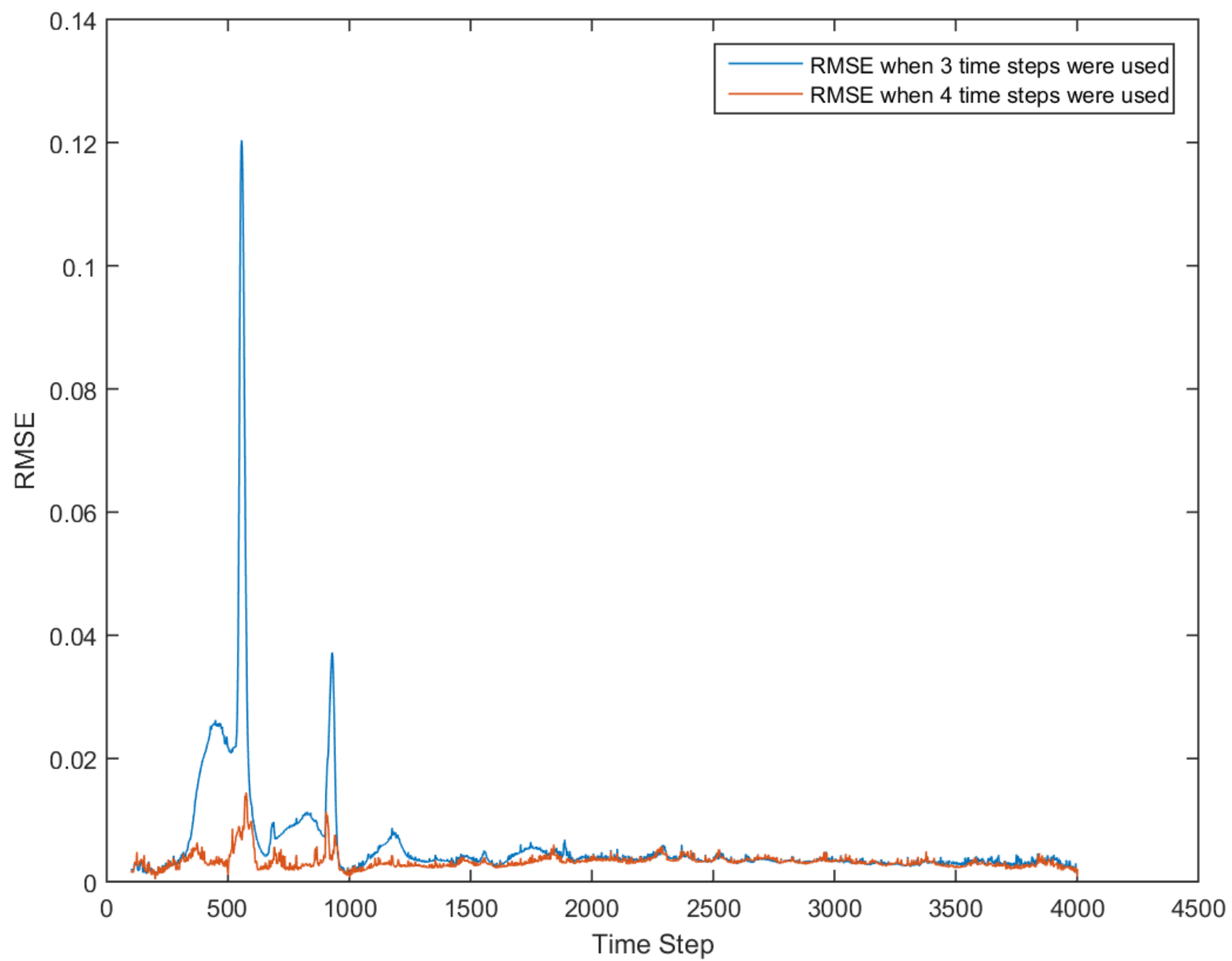
<i>Cell_{ijk} at time step (n + 1)</i>								
ε_g	P	P_s	u_g	v_g	w_g	u_s	v_s	w_s

Number of inputs $6 + 9 + 6 \times 9 = 69$

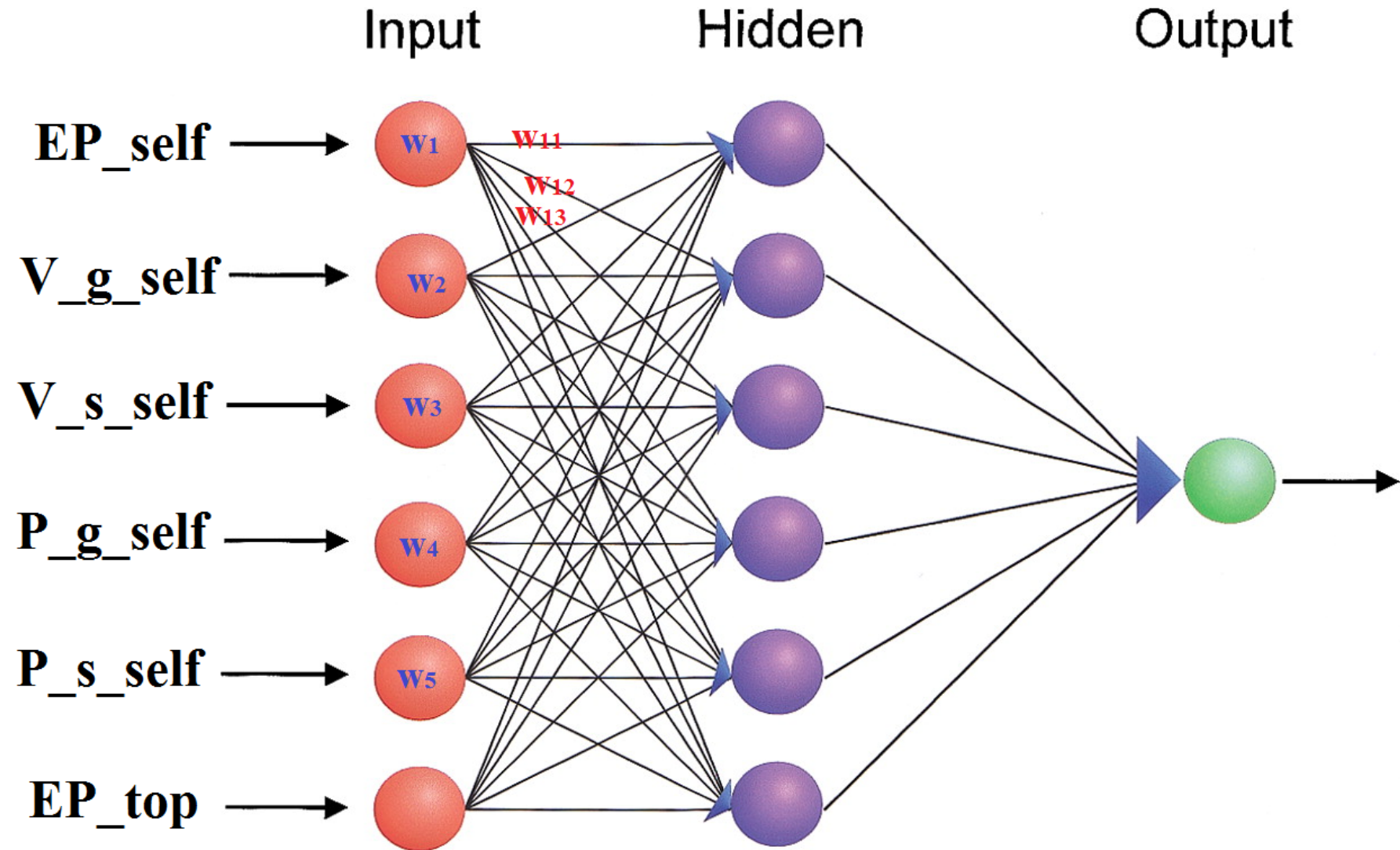
Number of outputs = 9

<i>Location of Cell_{ijk}</i>					
D_1	D_2	D_3	D_4	D_5	D_6



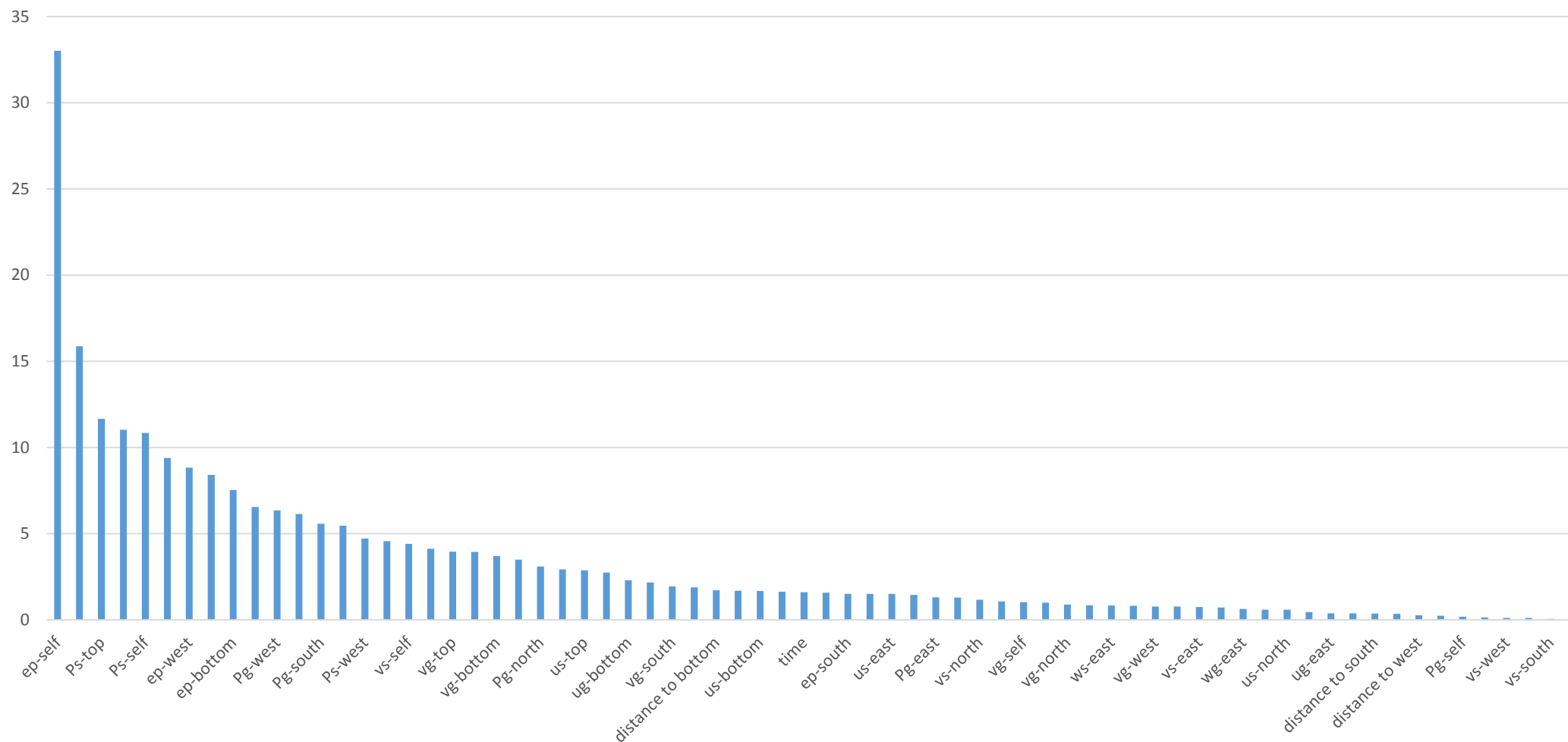


Optimize the number of inputs



Tornado chart by adding all the weights

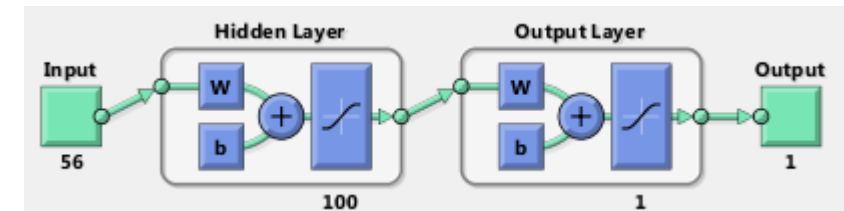
Tornado Chart of Neural Network Weights



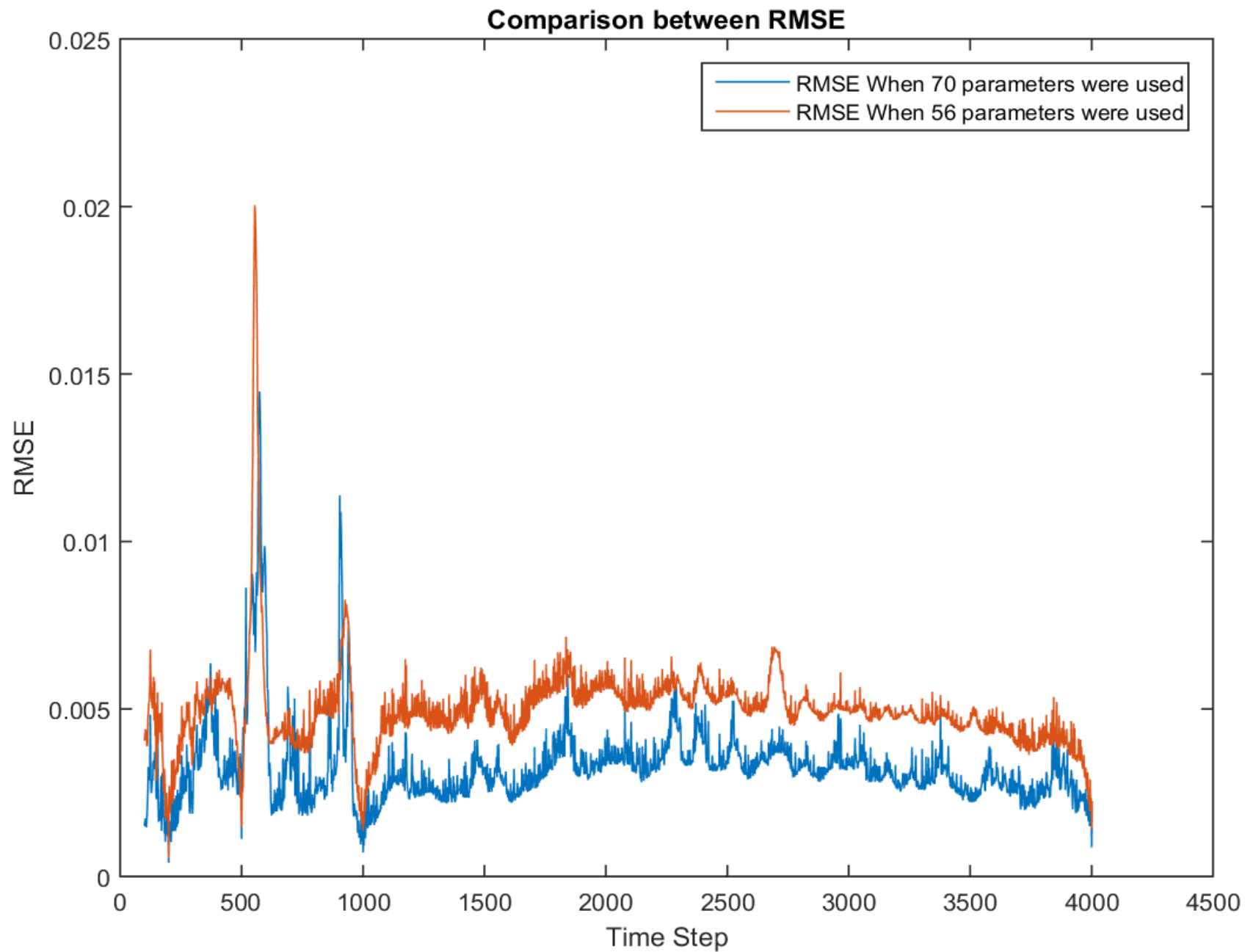
Eliminating 14 parameters

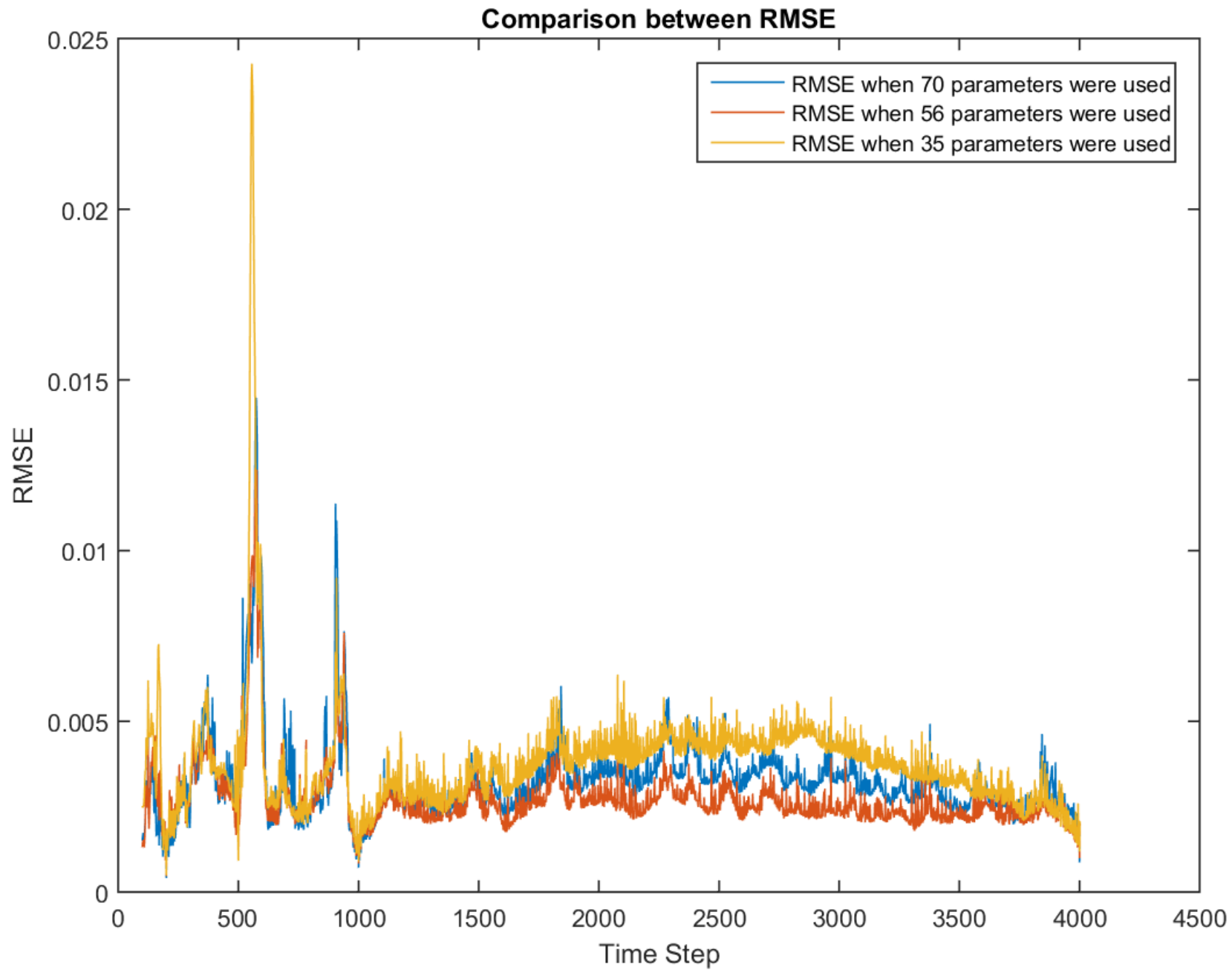
By parameter
ug-west
ug-north
ug-south
ug-east
vg-self
vg-west
wg-north
us-north
ws-west
vs-west
vs-bottom
distance to south
distance to east
distance to west

By location
vg-self
vs-bottom
ug-north
wg-north
us-north
ug-south
ws-west
vs-west
ug-west
vg-west
ug-east
distance to south
distance to east
distance to west

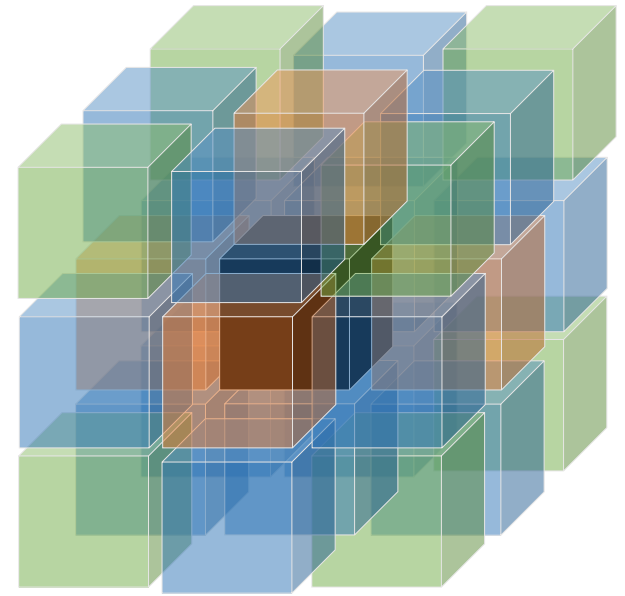
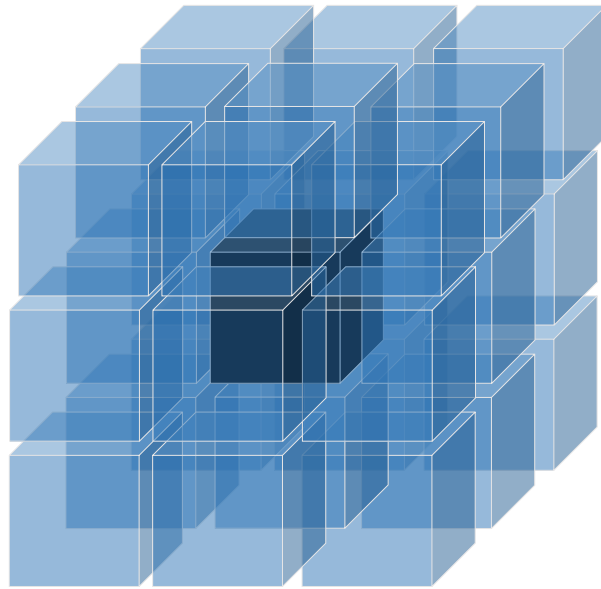
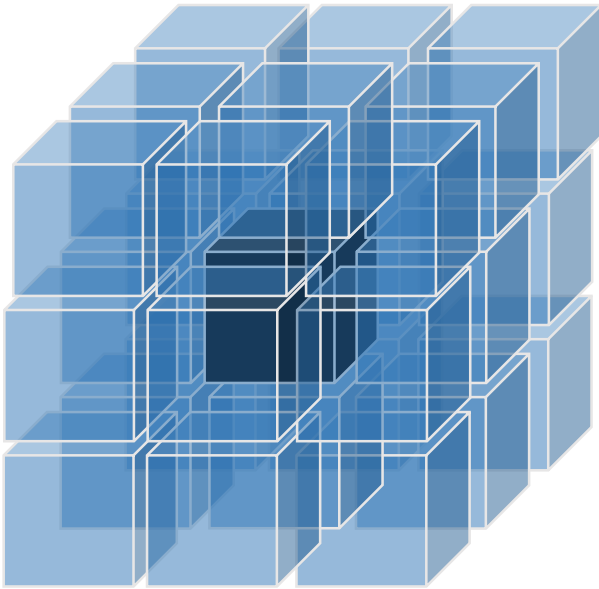


RSA

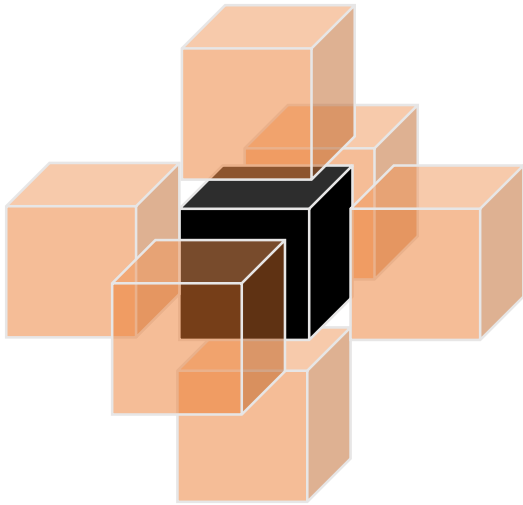




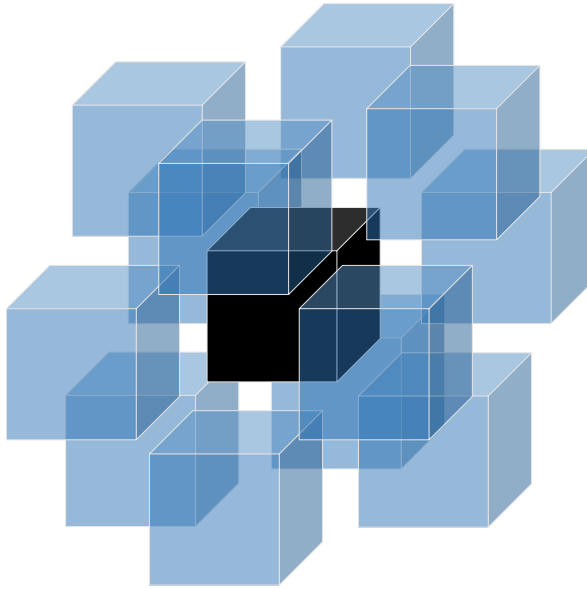
The Tier System



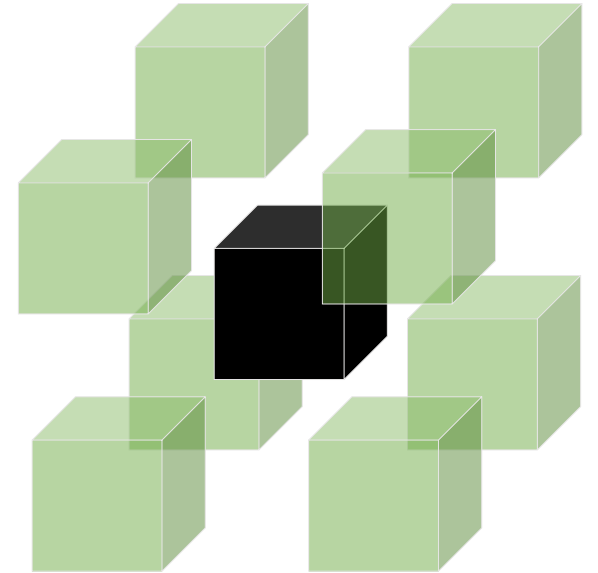
The Tier System



Tier One
Interface: Plane

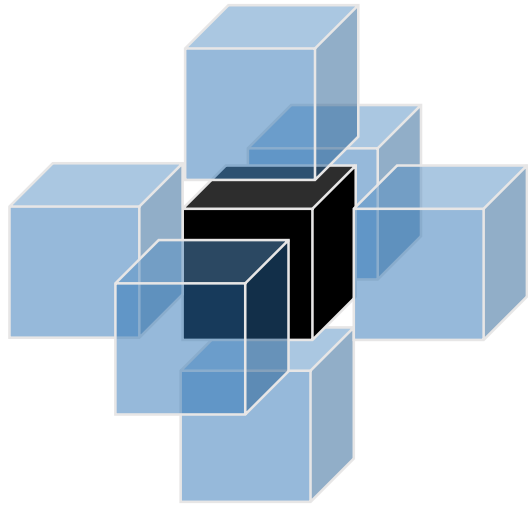


Tier Two
Interface: Line

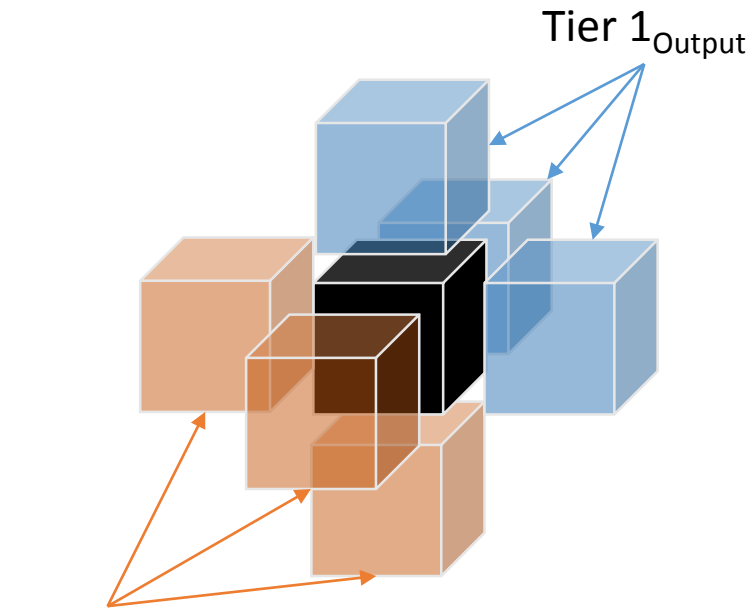


Tier Three
Interface: Point

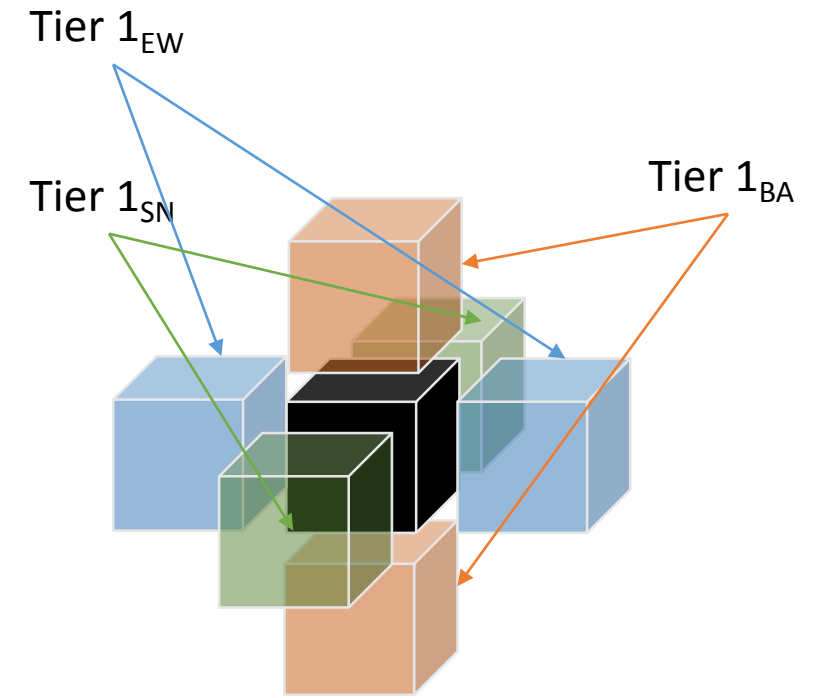
Tier One - Subsystem



Tier One
Interface: Plane



Tier One
Two Subsystems
Tier 1_{Input} and Tier 1_{Output}



Tier One
Three Subsystems
Tier 1_{EW} , Tier 1_{SN} , and Tier 1_{BA}