

Combining SU2 and Graph Neural Networks for Fluid Flow Prediction

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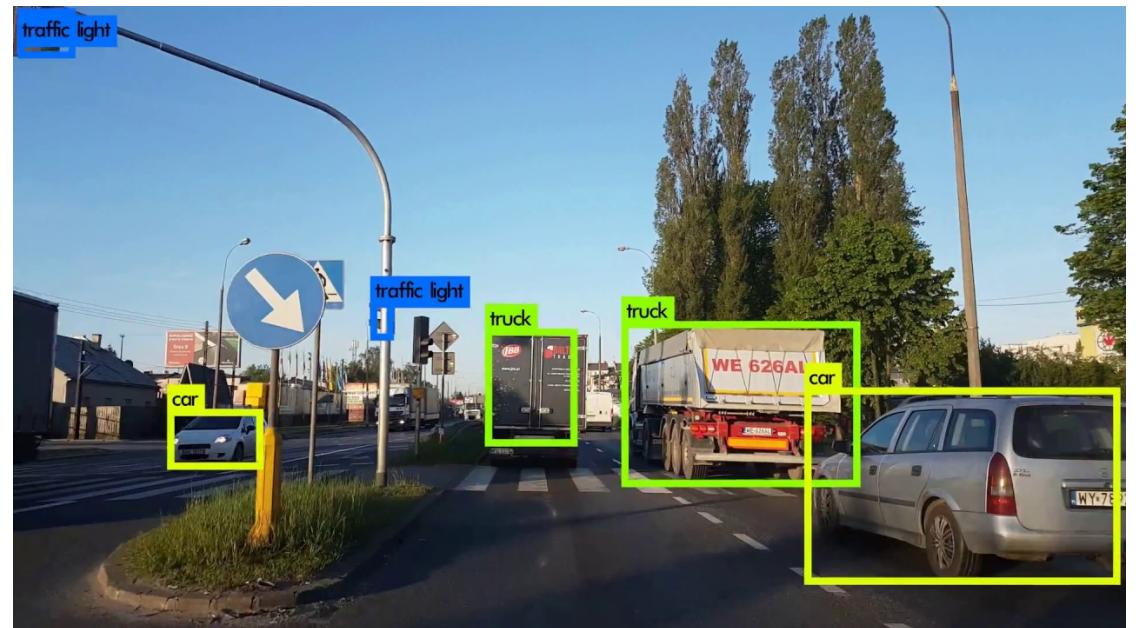
SU2
foundation



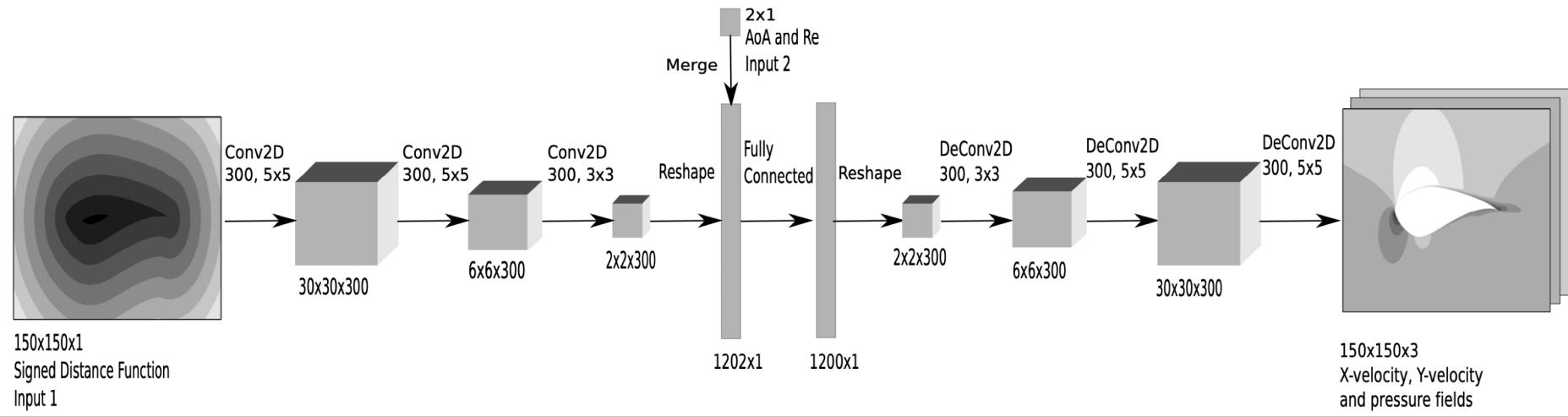
BOSCH

Successes of Deep Learning

Learning complex input-output relations

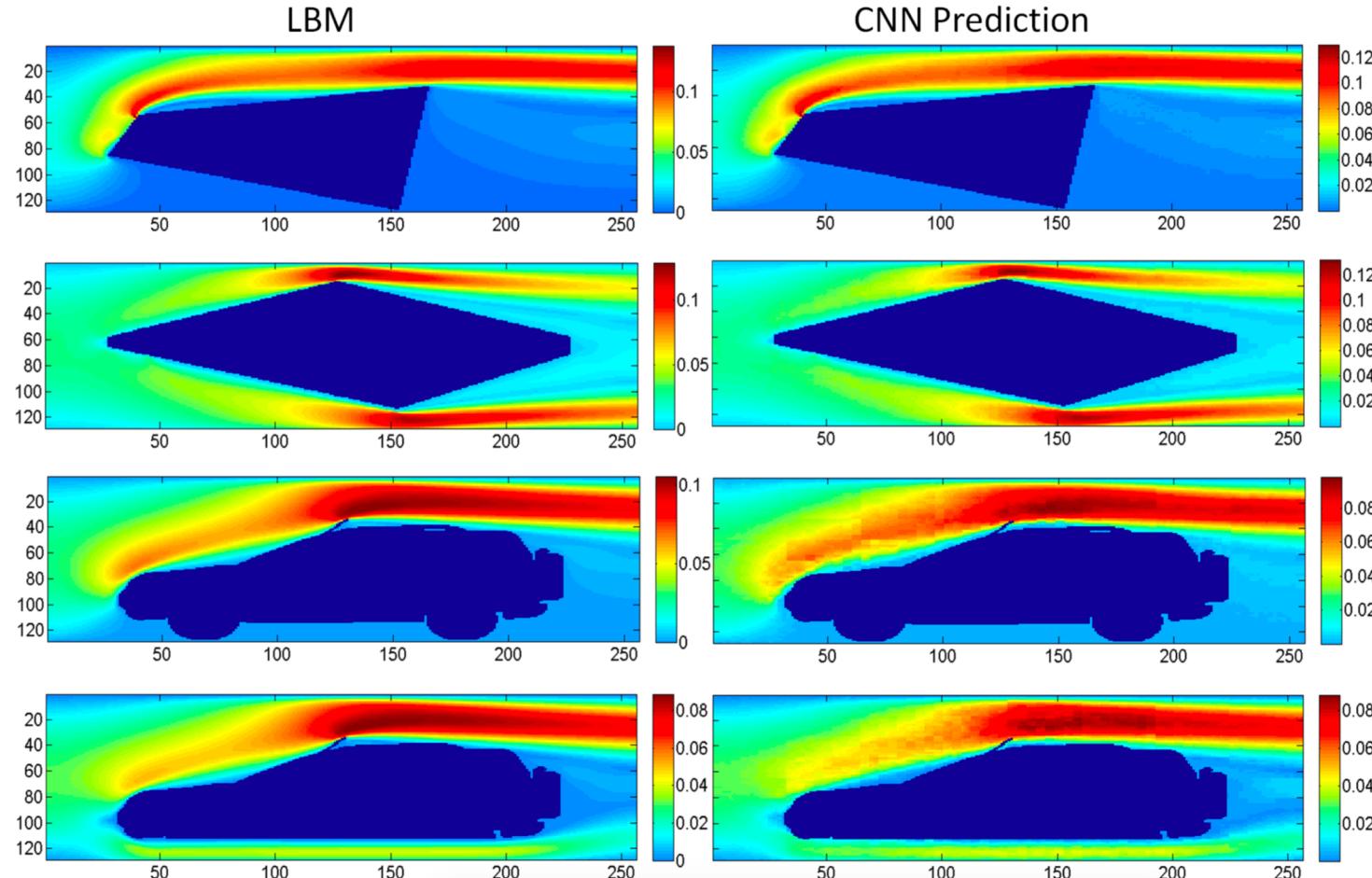


Machine learning and CFD



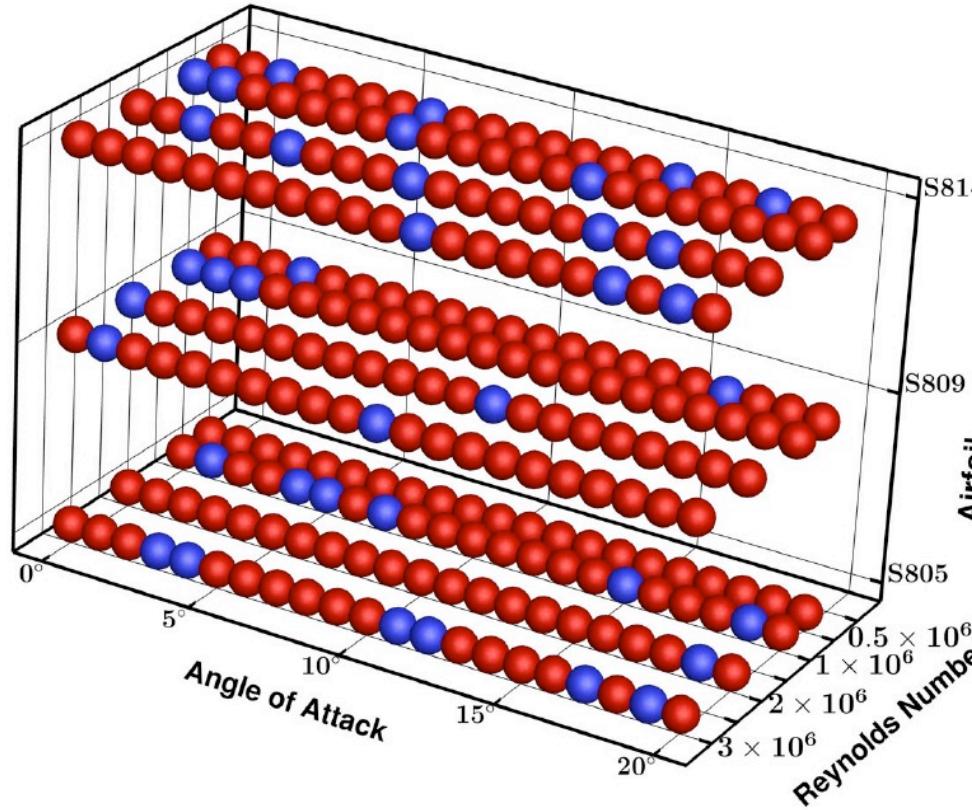
Machine learning and CFD

Standard Convolutional networks can only be applied to regular grids



Machine learning and CFD

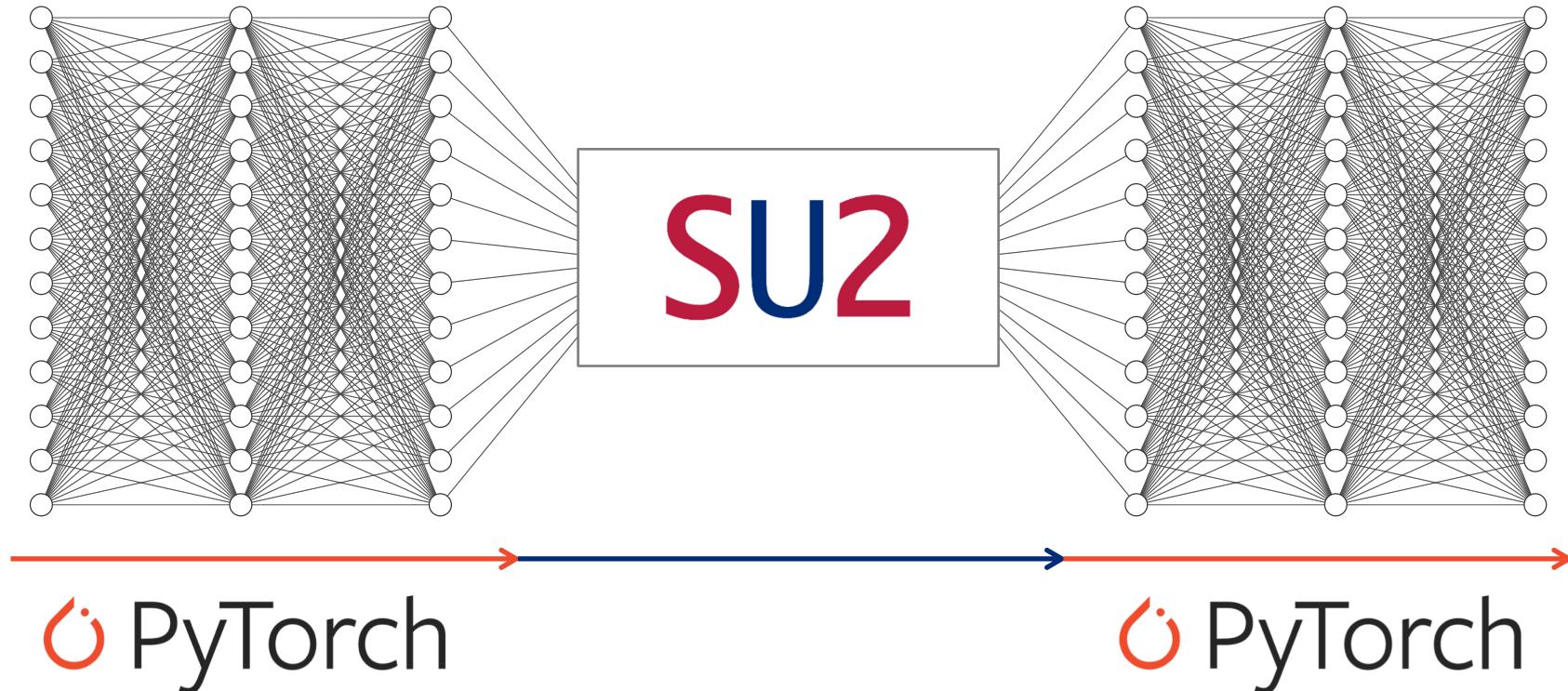
Works well on data within the distribution seen during training



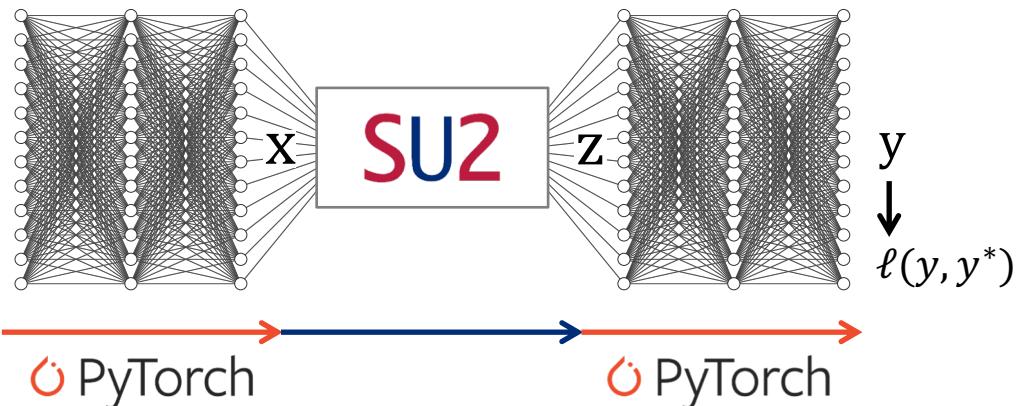
Combining SU2 and Graph Networks

- Operate directly on non-uniform meshes using graph networks
- Embed differentiable CFD solver into deep learning model for better generalization

Schematic pipeline



Interface overview



- Config file

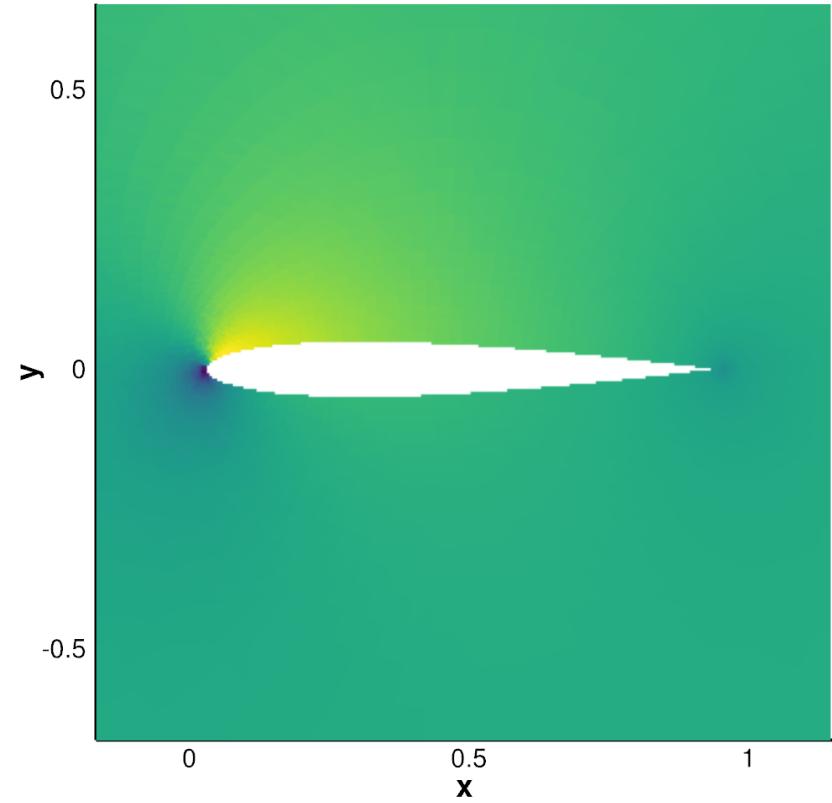
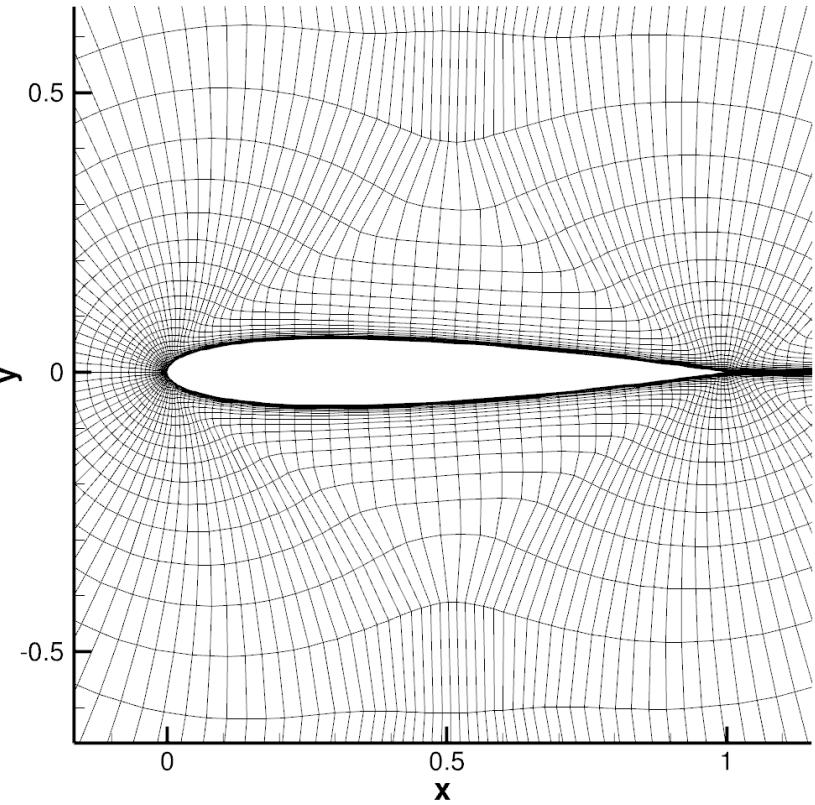
```
% Regular config file...

DIFF_INPUTS = MESH_X, MESH_Y
DIFF_OUTPUTS = VEL_X, VEL_Y, PRESS
```

- Python

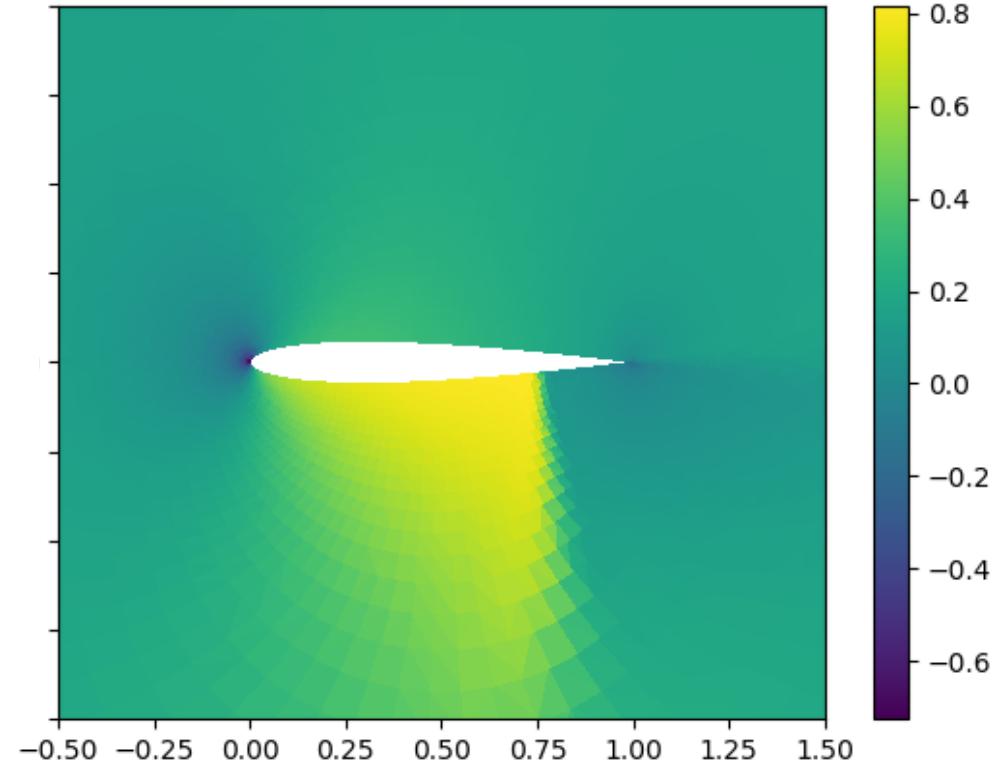
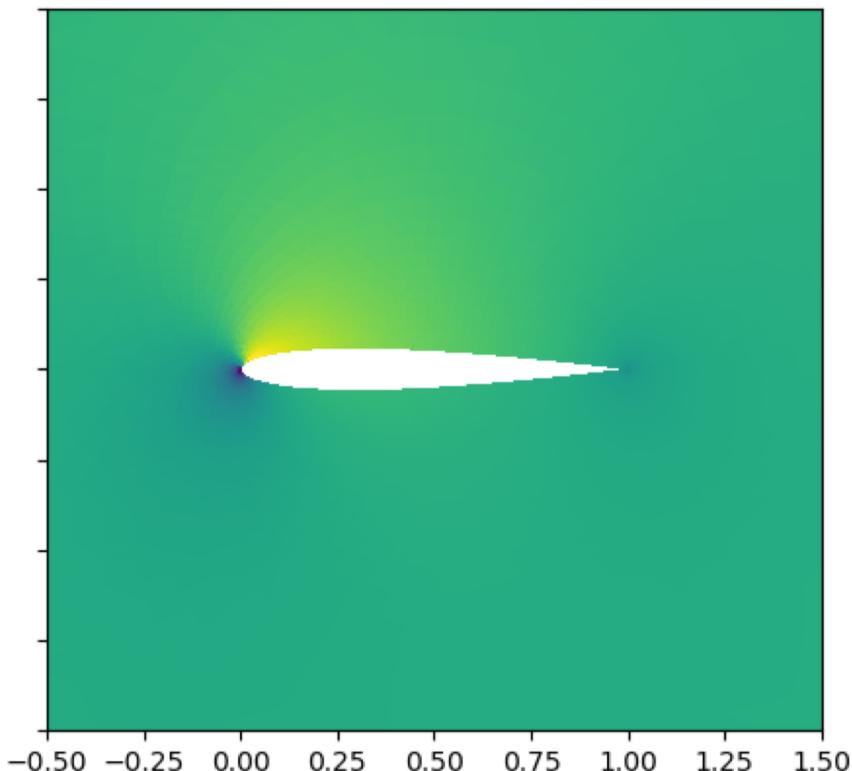
```
import torch
from su2_function import SU2Function
# process some data in torch
x = torch.operation(data)
# input into SU2 and run simulation
su2 = SU2Function(config_file)
z = su2(x)
# process more in torch
y = torch.operation(z)
# set loss and get gradients
loss = torch.nn.MSELoss(y, y_star)
loss.backward() # dloss/dx available
```

Simulating flow around an airfoil

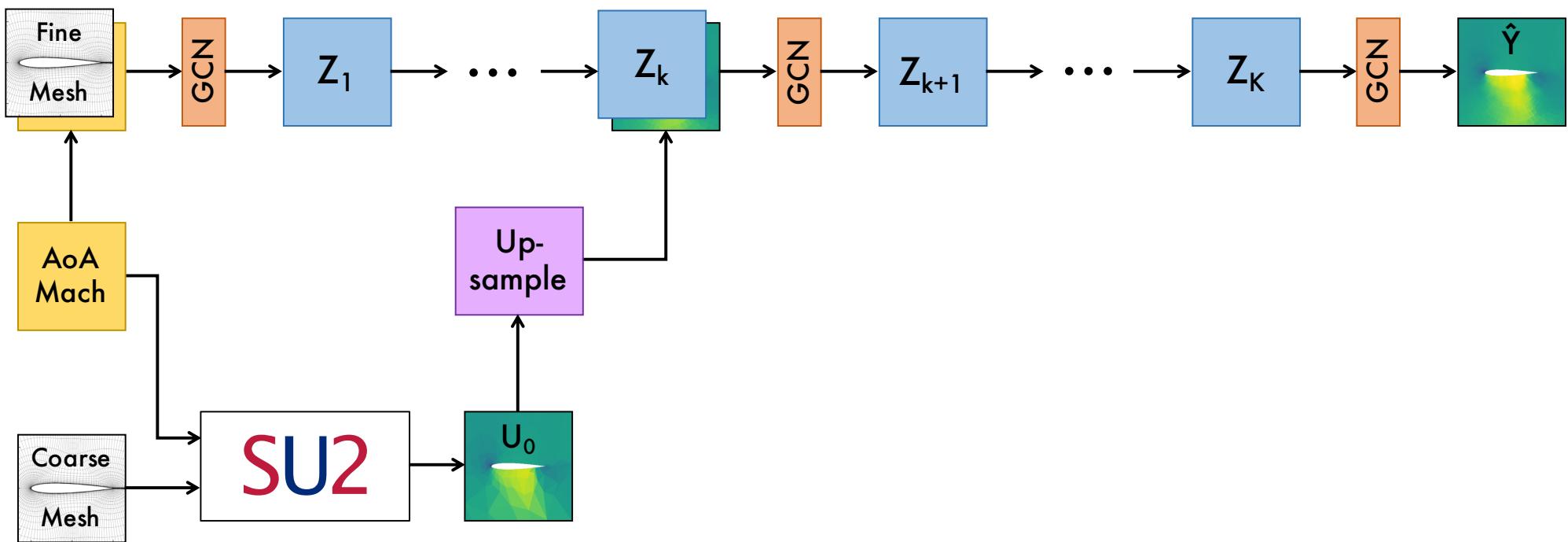


Physical Parameters

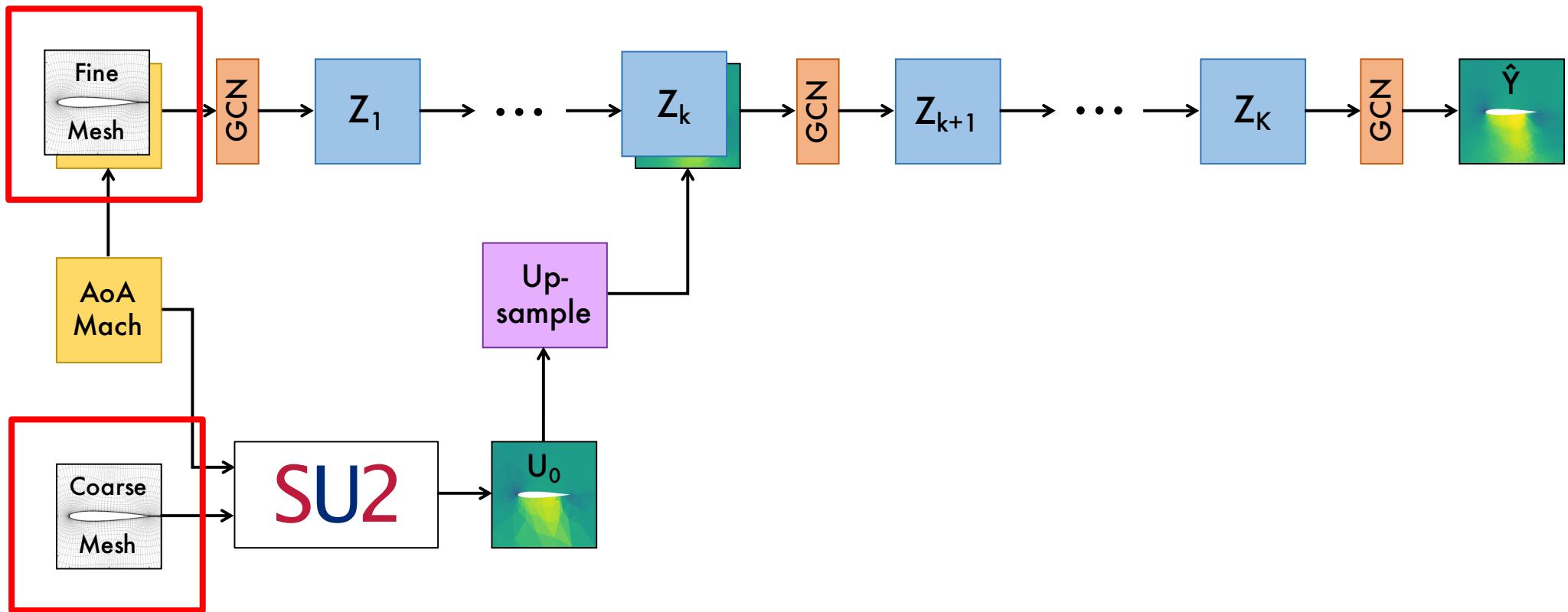
$\text{AoA} \in \{-10, -9, \dots, 9, 10\}$
 $\text{Mach} \in [0.2, 0.8]$



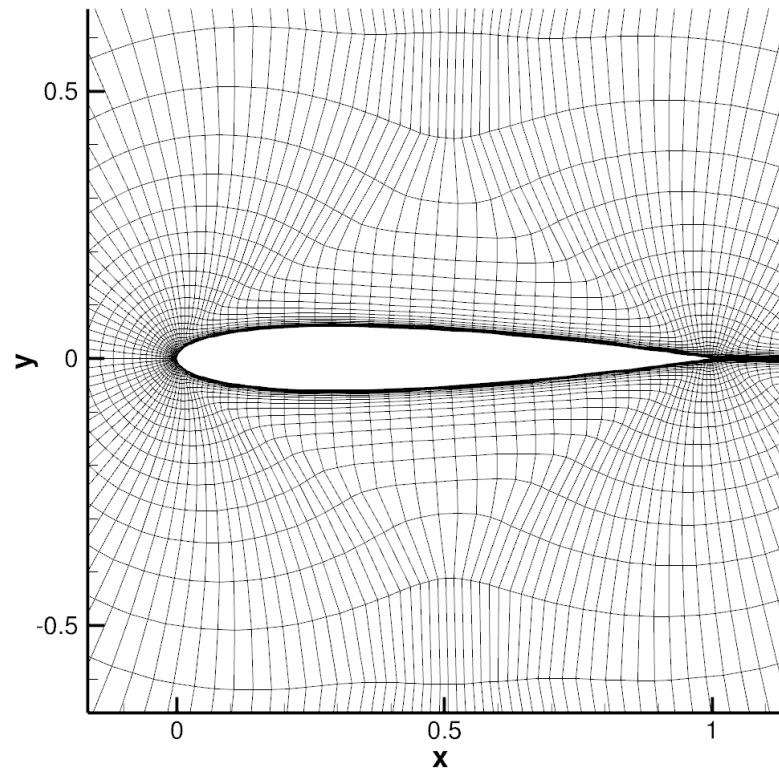
CFD-GCN



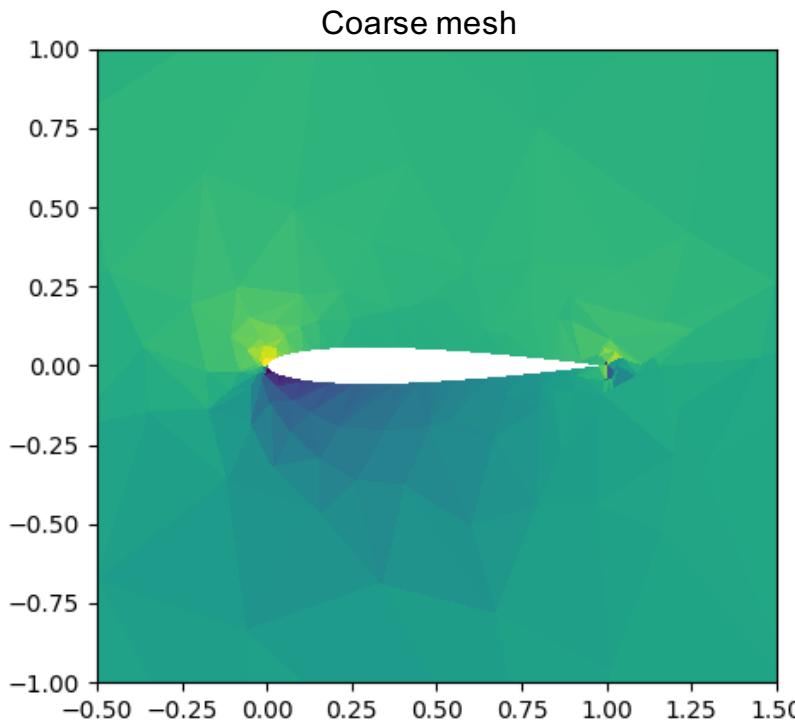
CFD-GCN



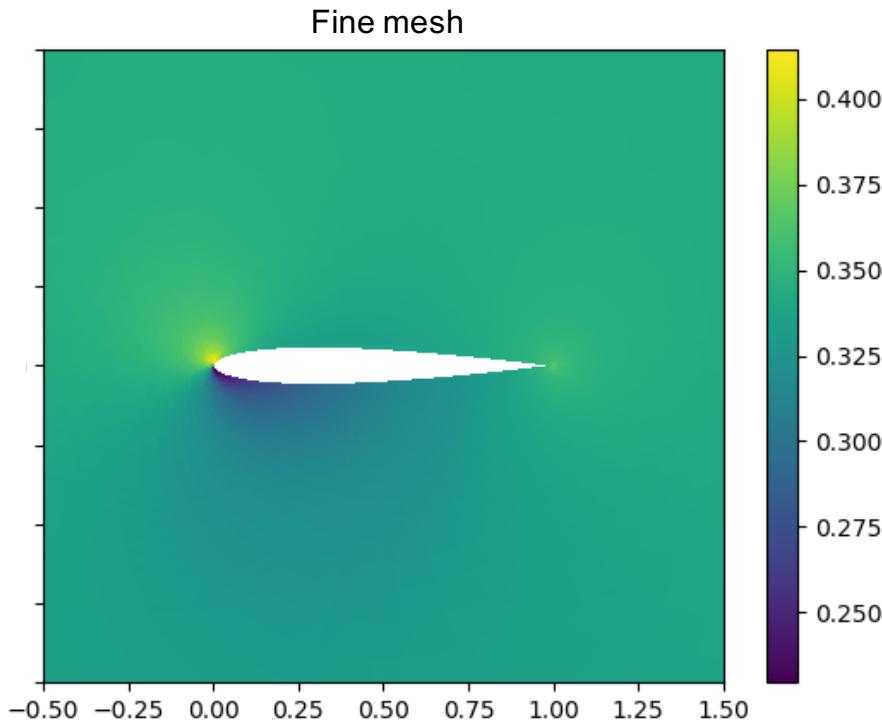
Airfoil Mesh



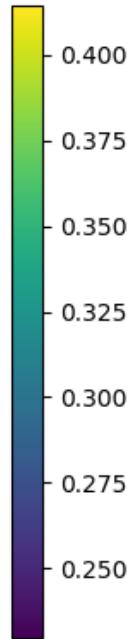
Coarse and Fine Meshes



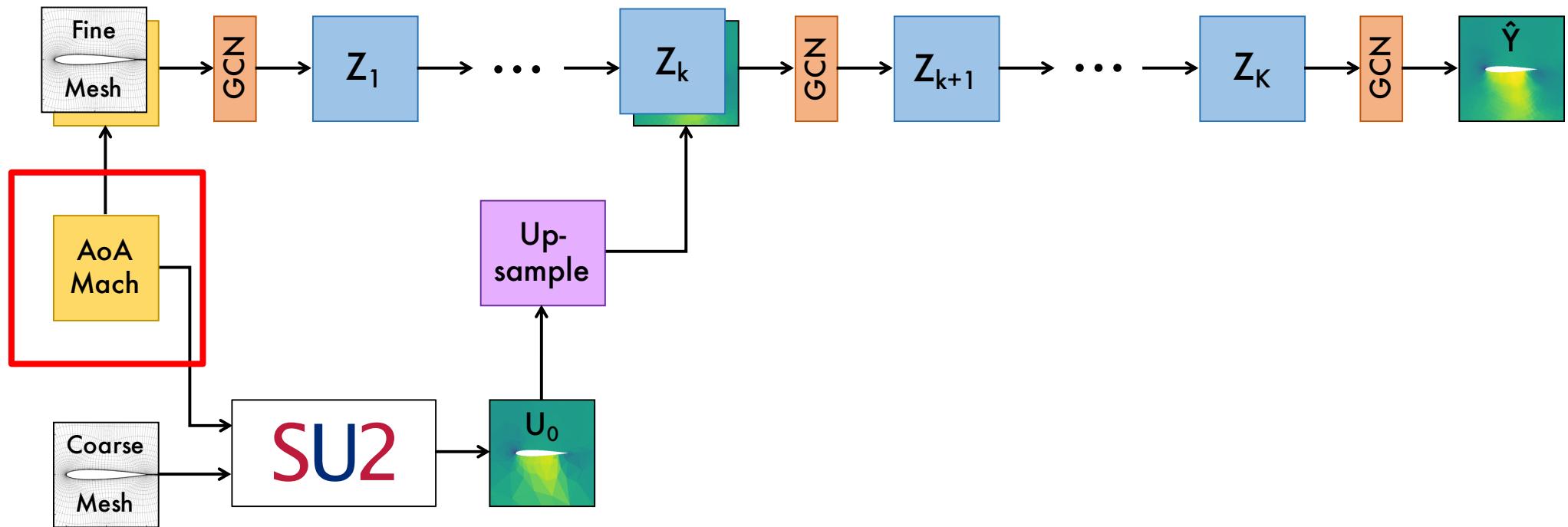
≈ 300 nodes



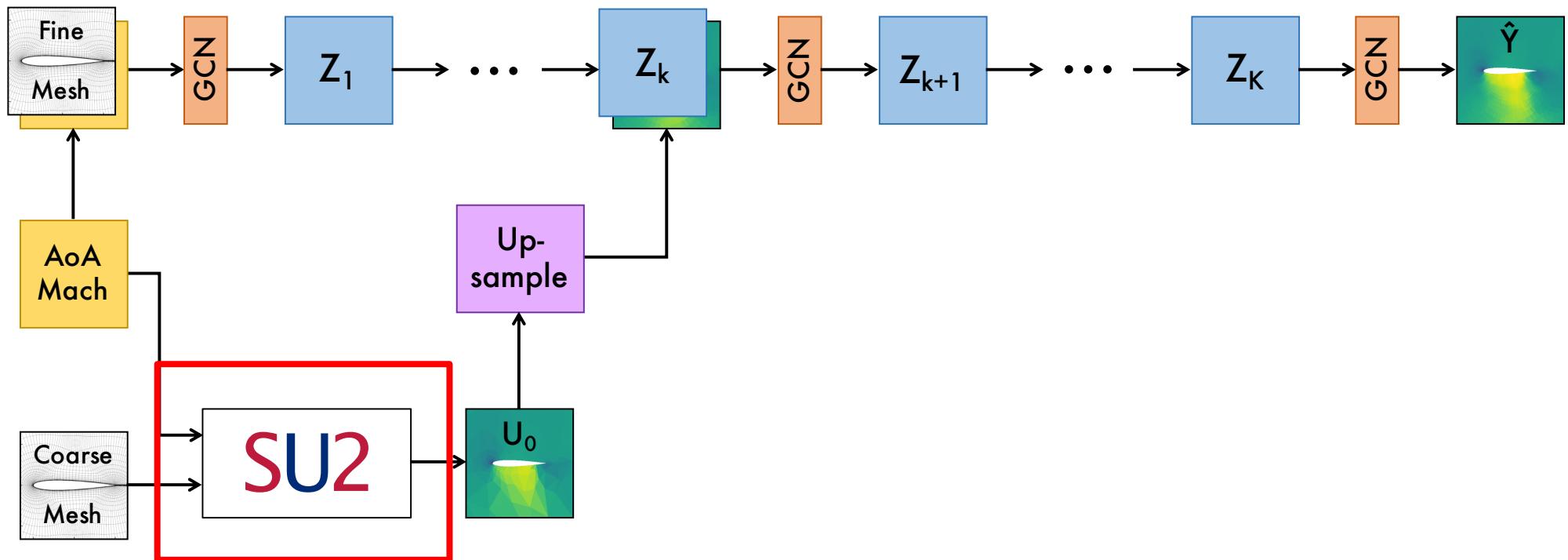
≈ 6000 nodes



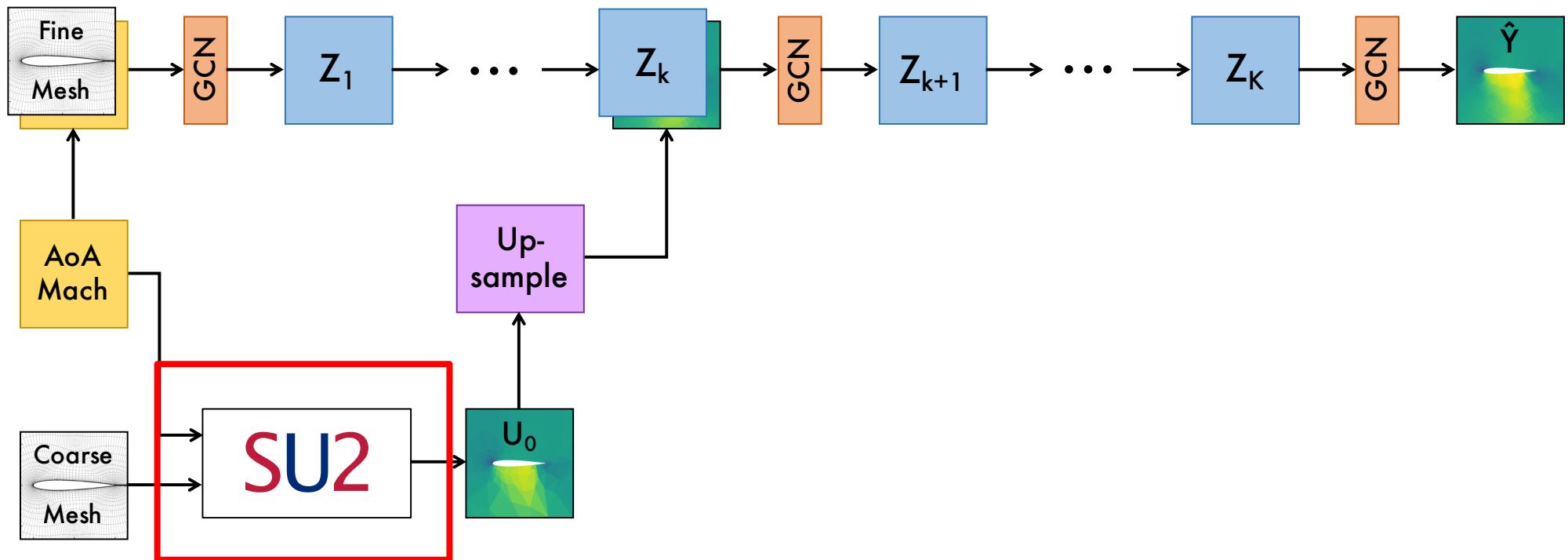
CFD-GCN



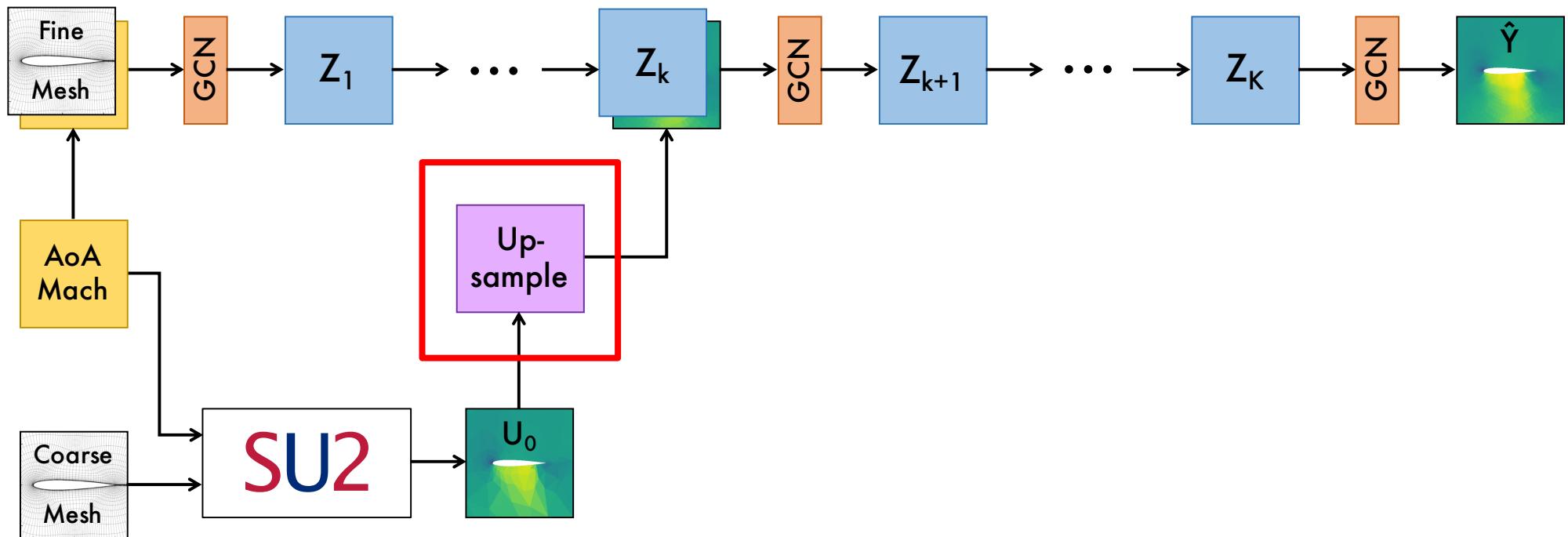
CFD-GCN



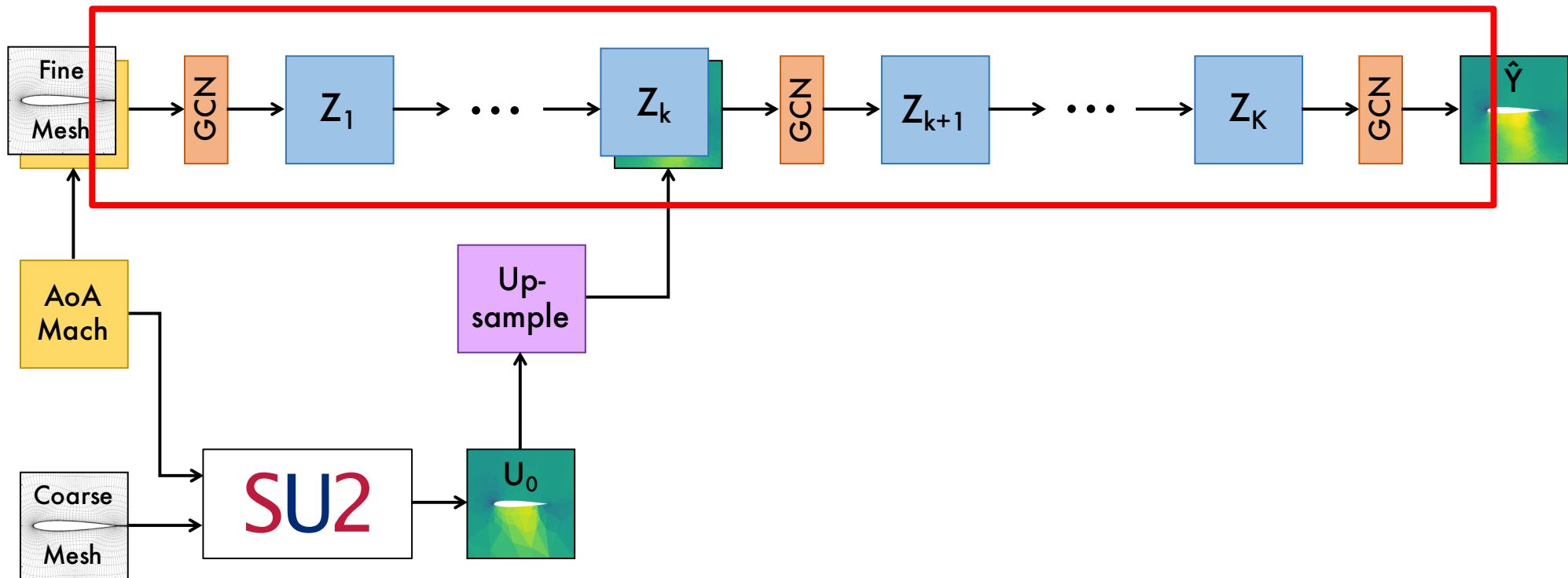
CFD-GCN



CFD-GCN



CFD-GCN

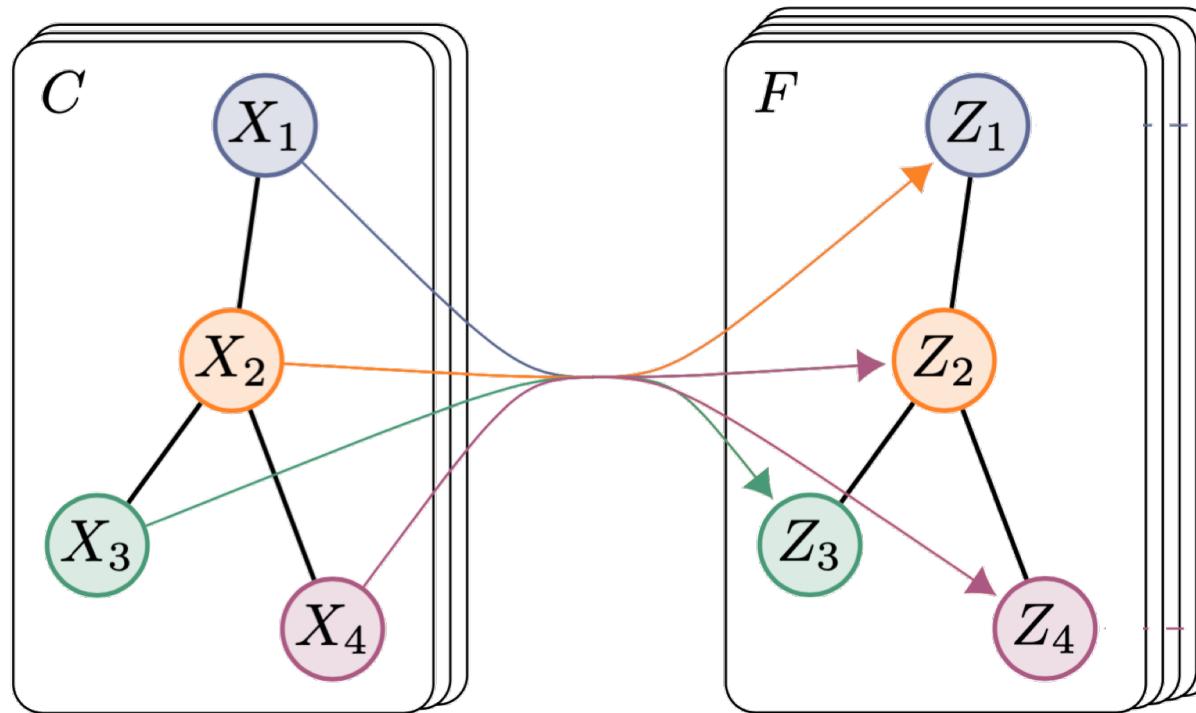


GCN Layer

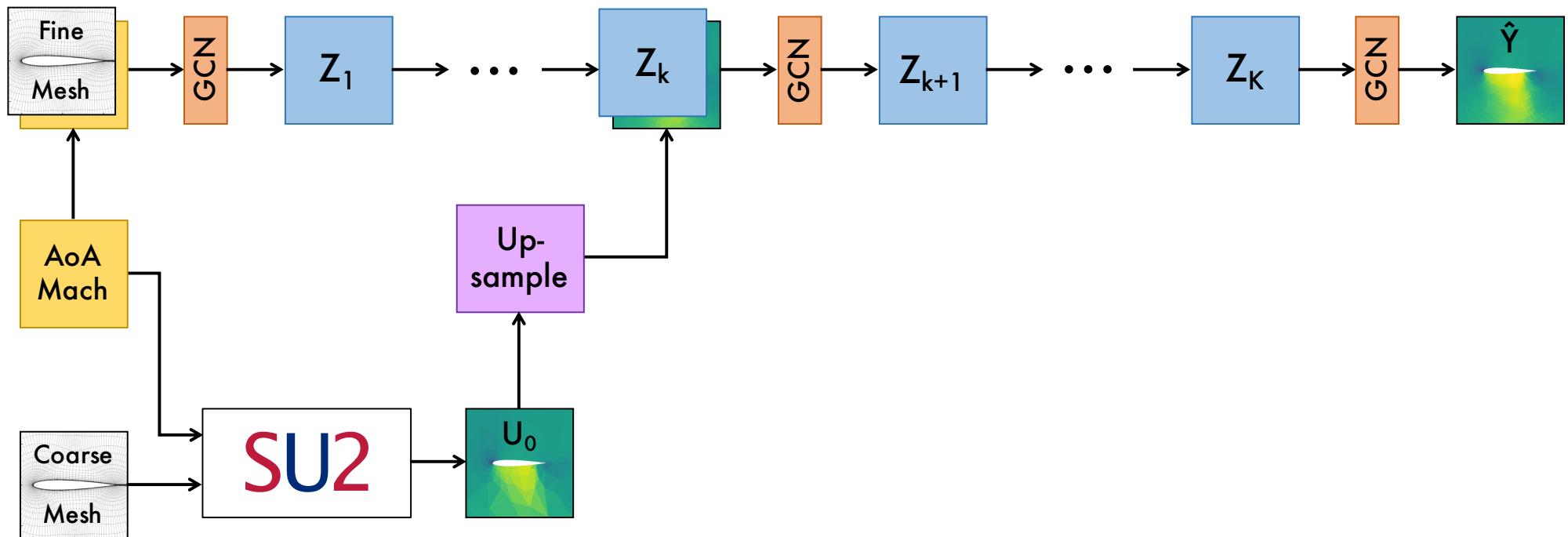
$$Z = \text{GCN}(X) := \tilde{A} X W + b$$

$$X \in \mathbb{R}^{N \times C}$$

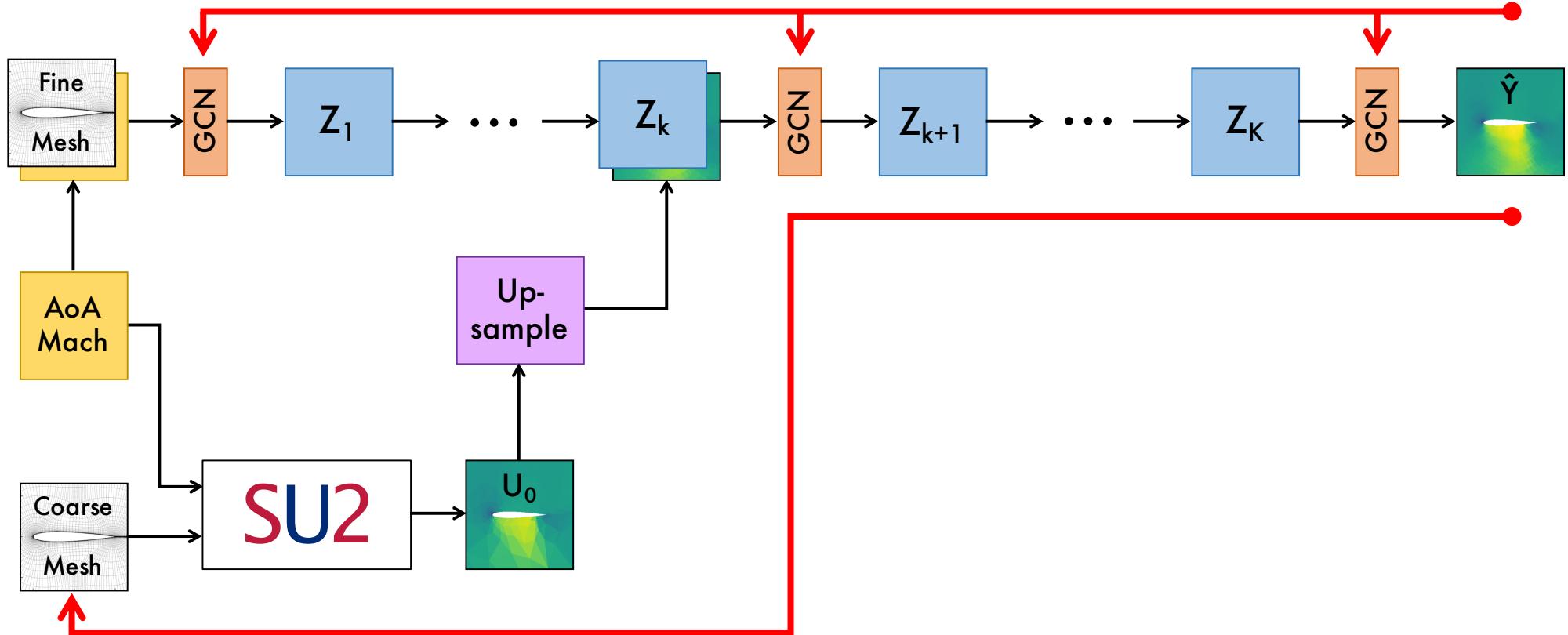
$$Z \in \mathbb{R}^{N \times F}$$



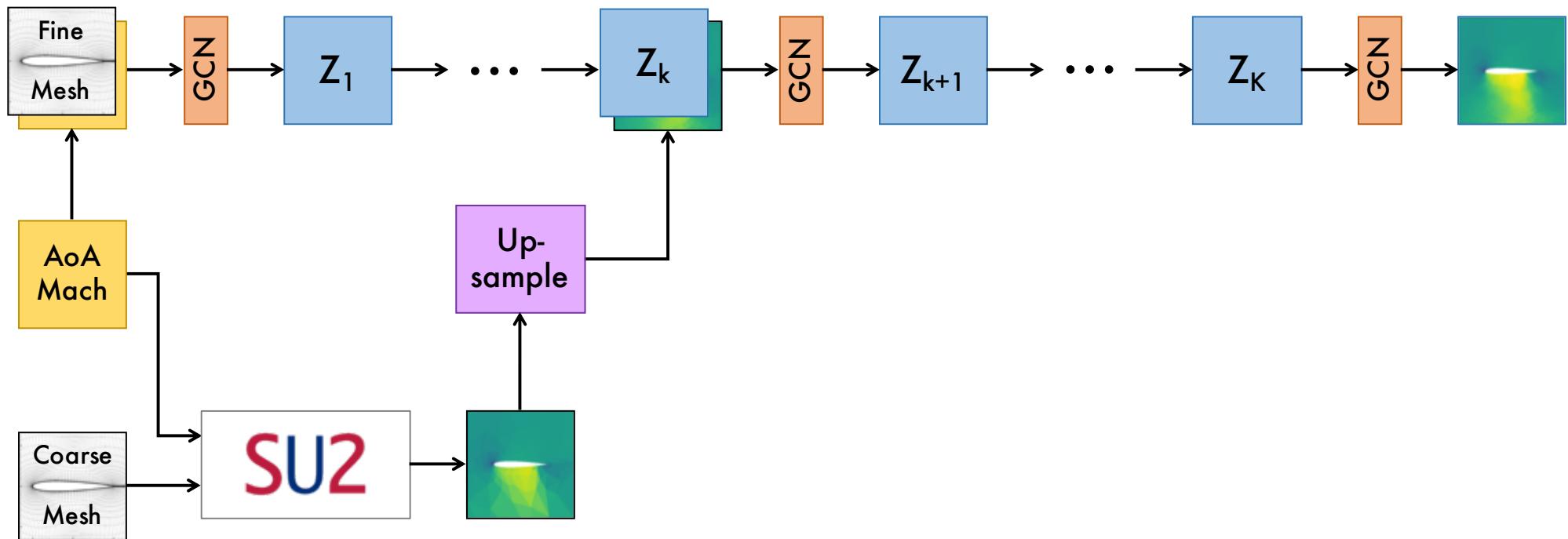
CFD-GCN



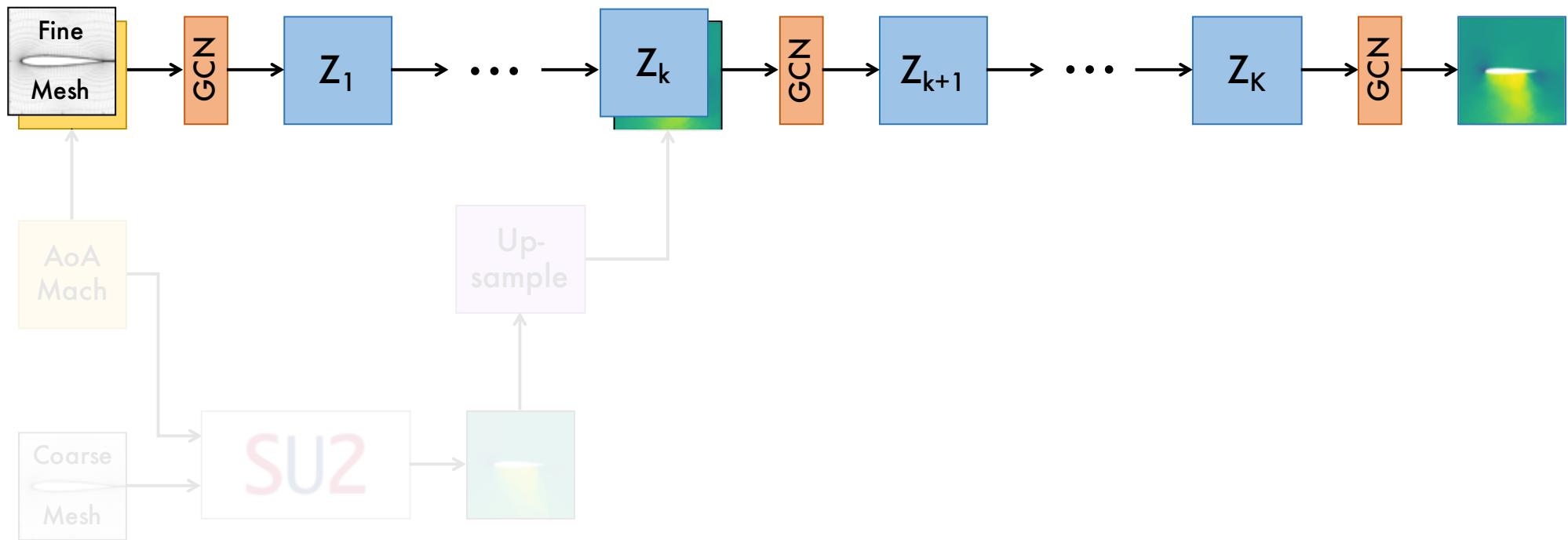
CFD-GCN Backpropagation



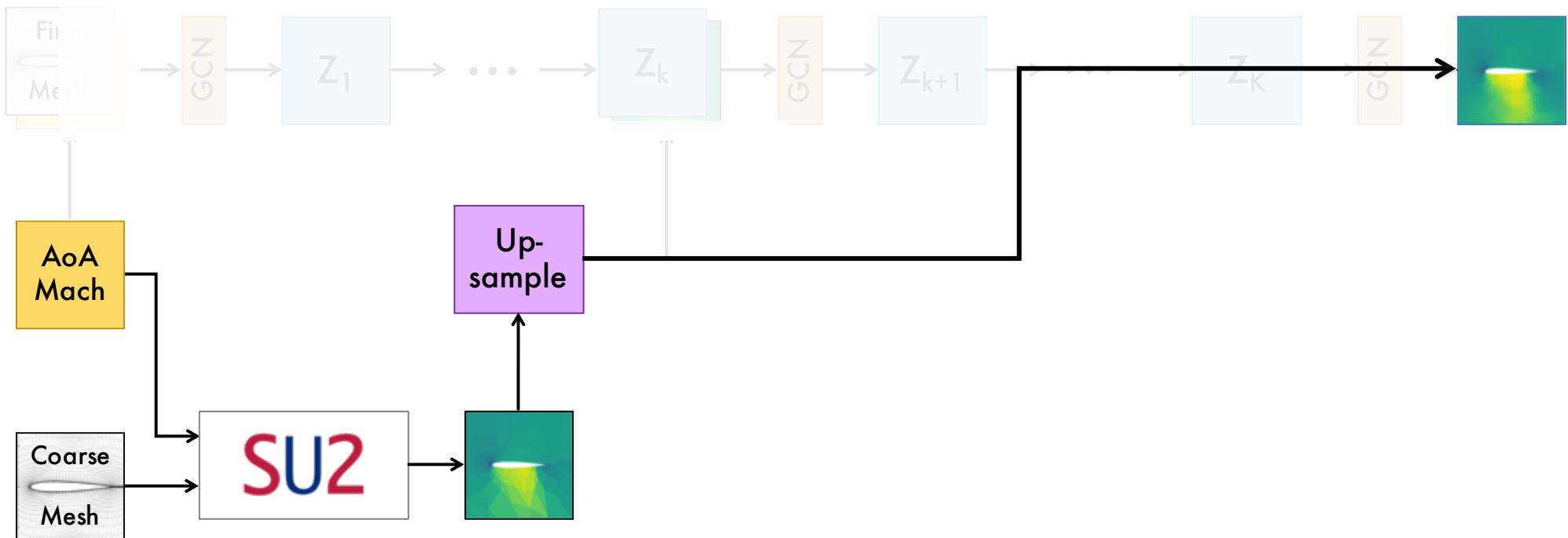
CFD-GCN



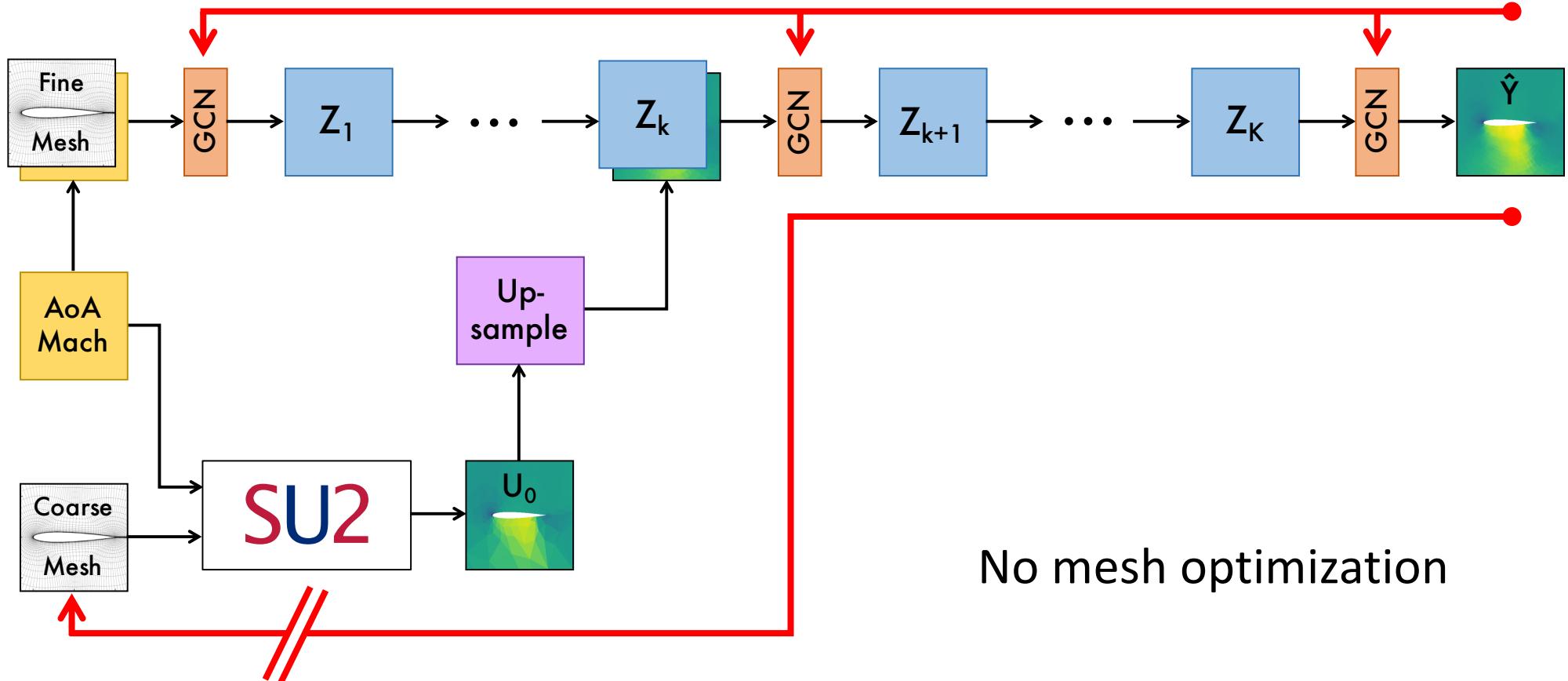
GCN Baseline



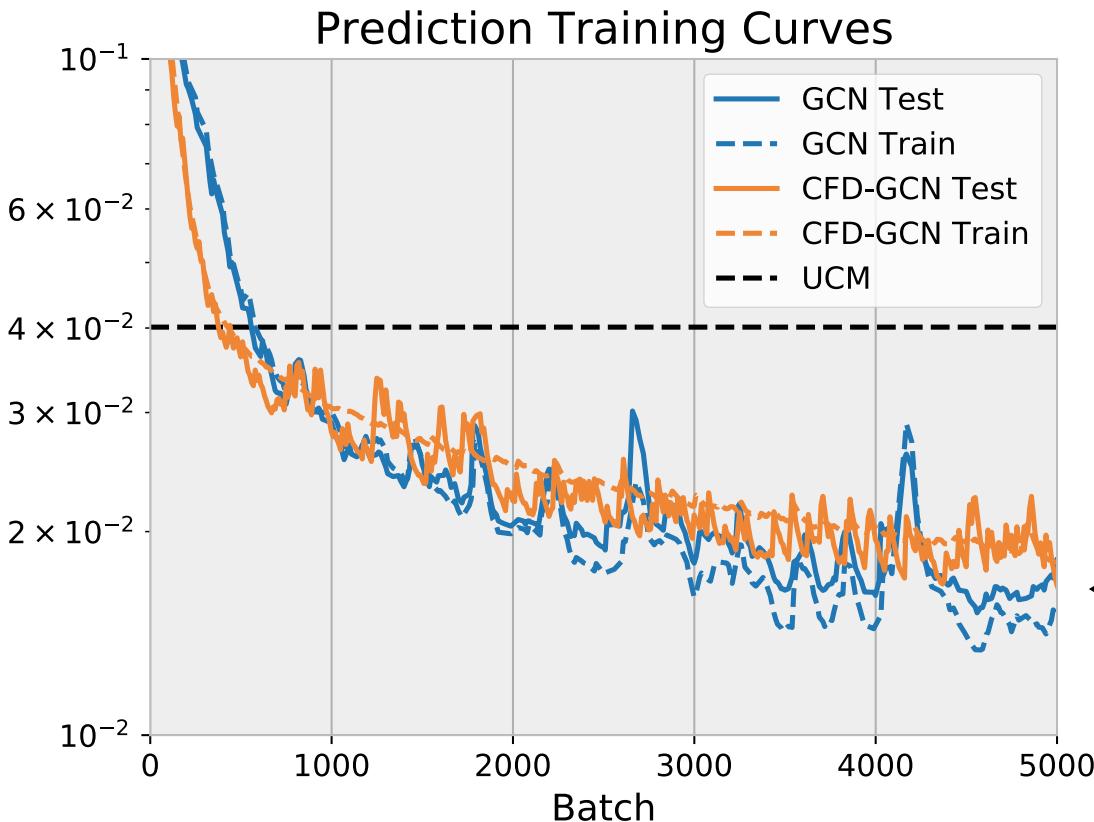
Upsampled Coarse Model (UCM) Baseline



Frozen Mesh Baseline

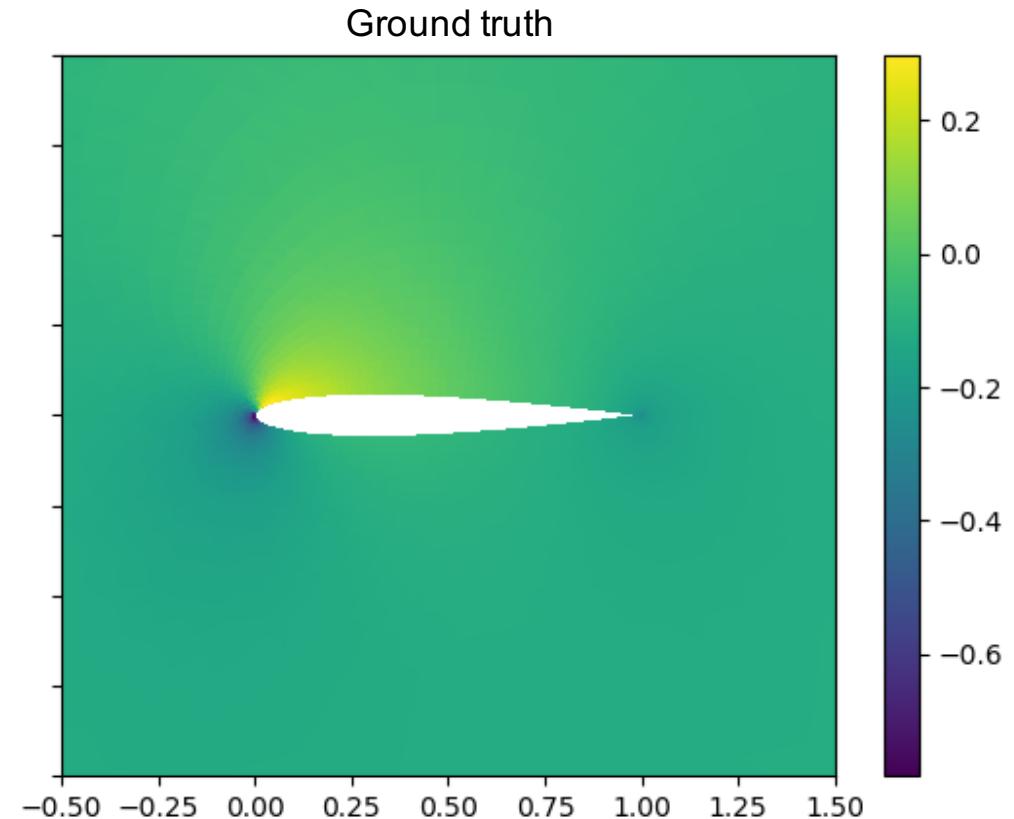
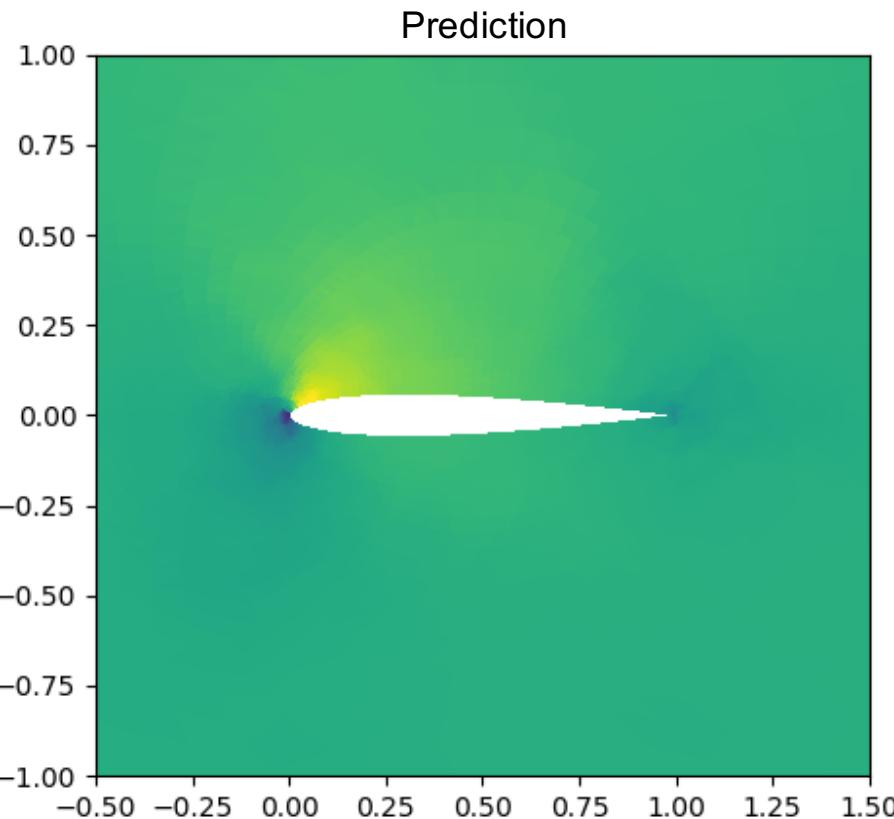


Prediction Results

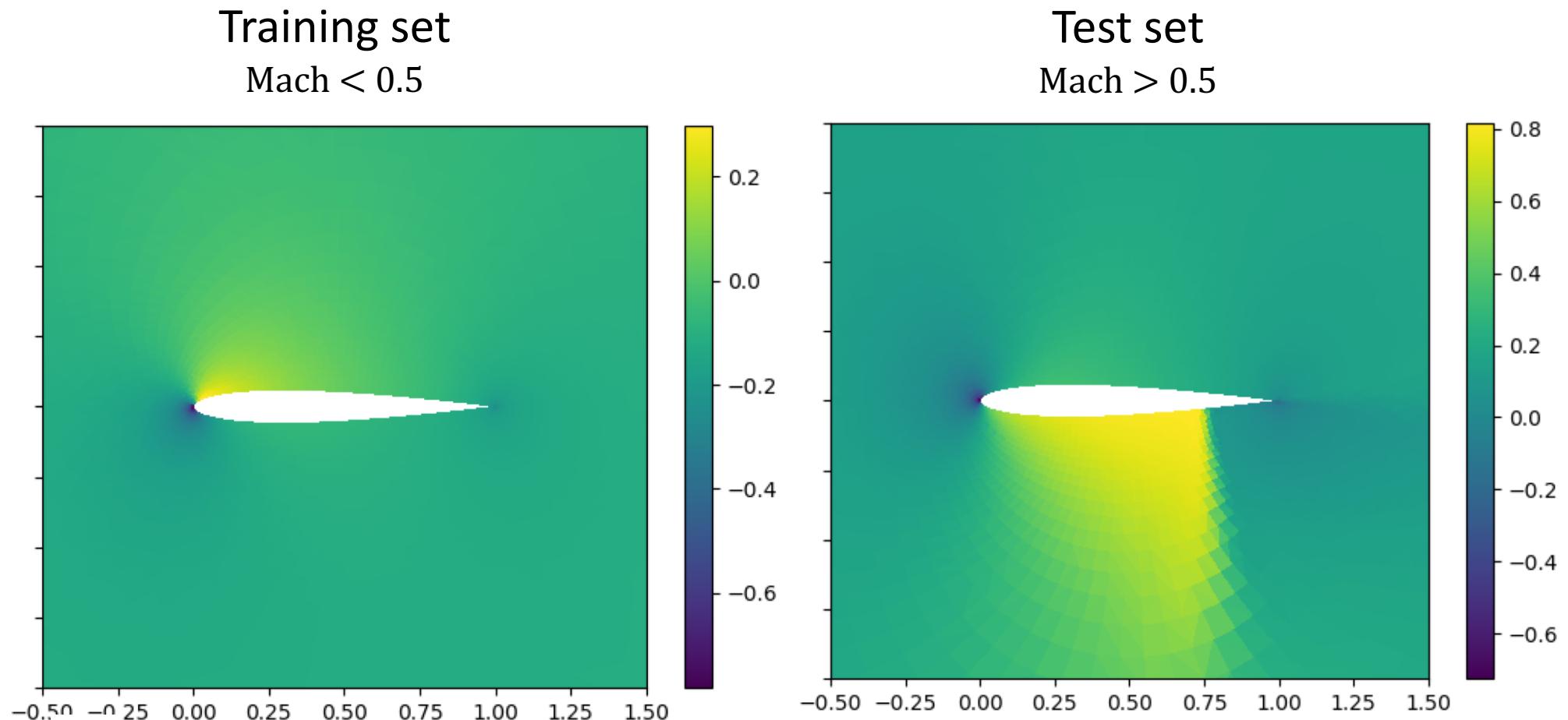


Similar performance between
GCN and CFD-GCN, with slight
edge to GCN due to overfitting

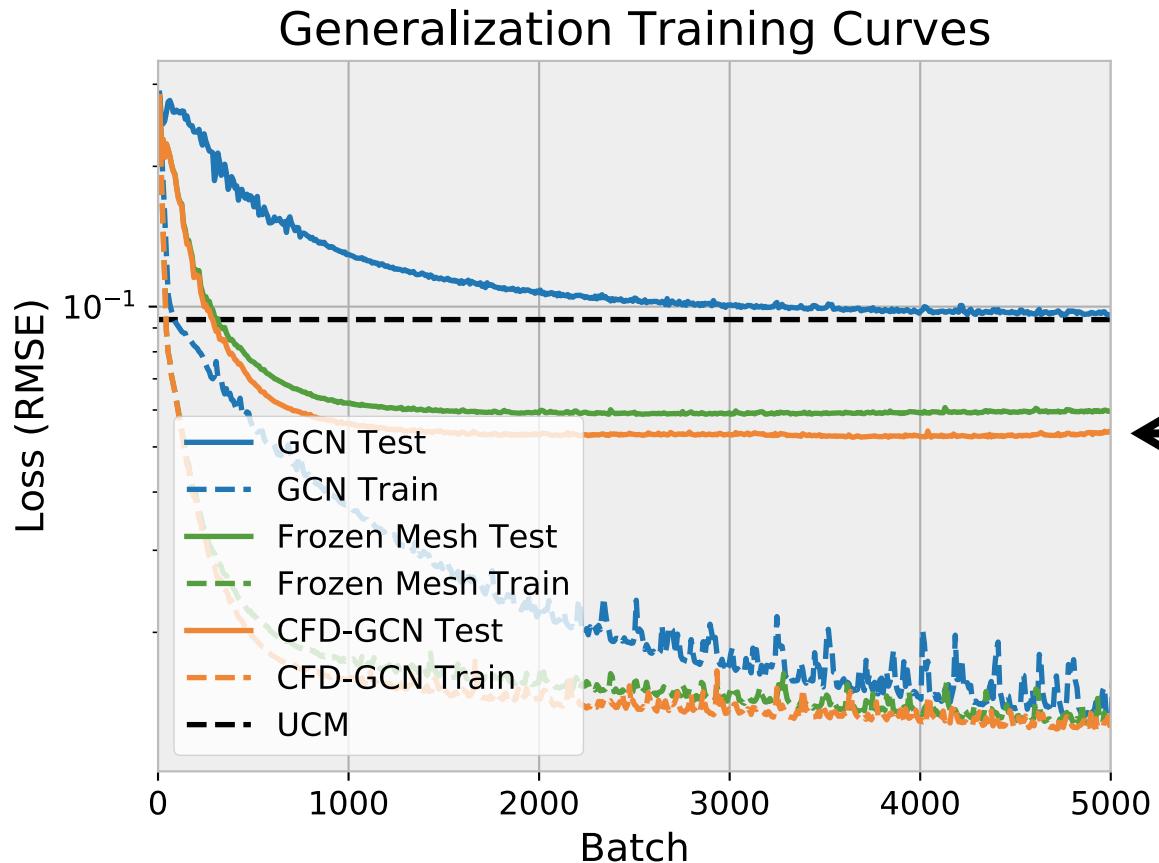
Prediction Results



Generalization Task



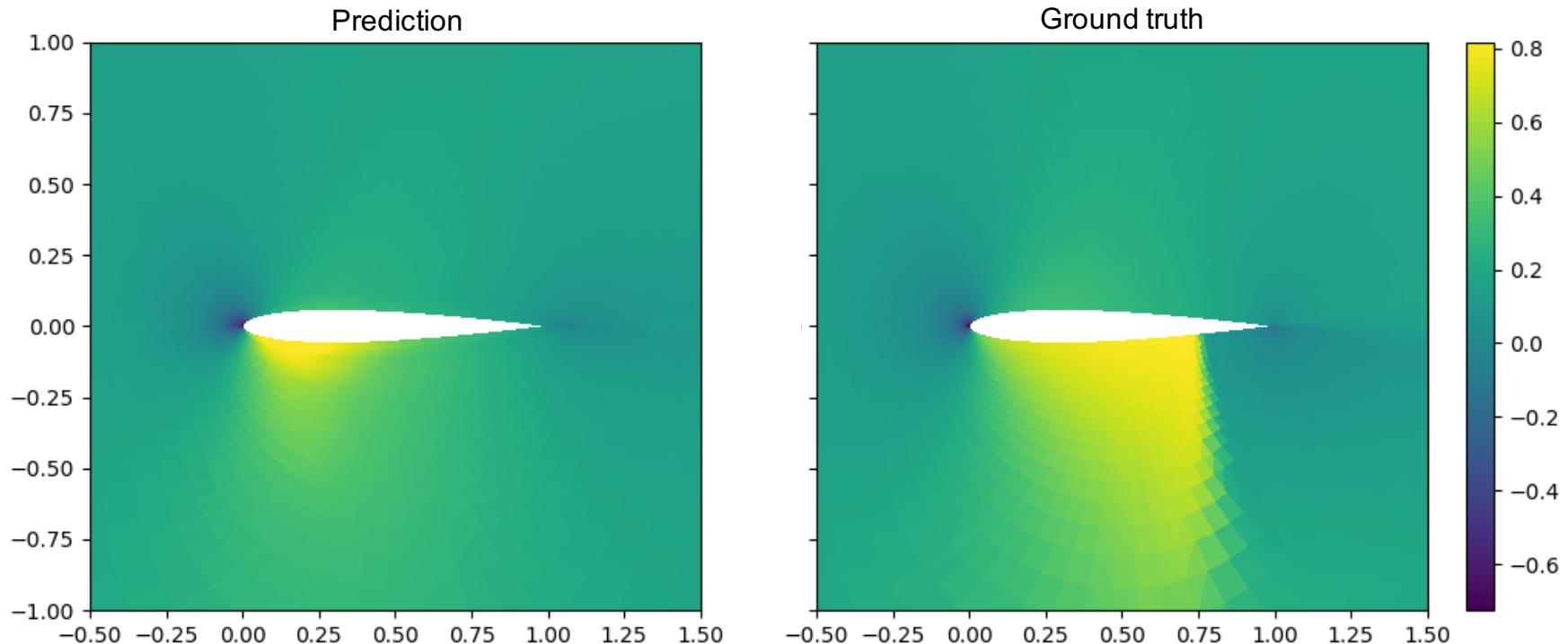
Generalization Results



CFD-GCN outperforms all baselines when generalizing to unseen conditions

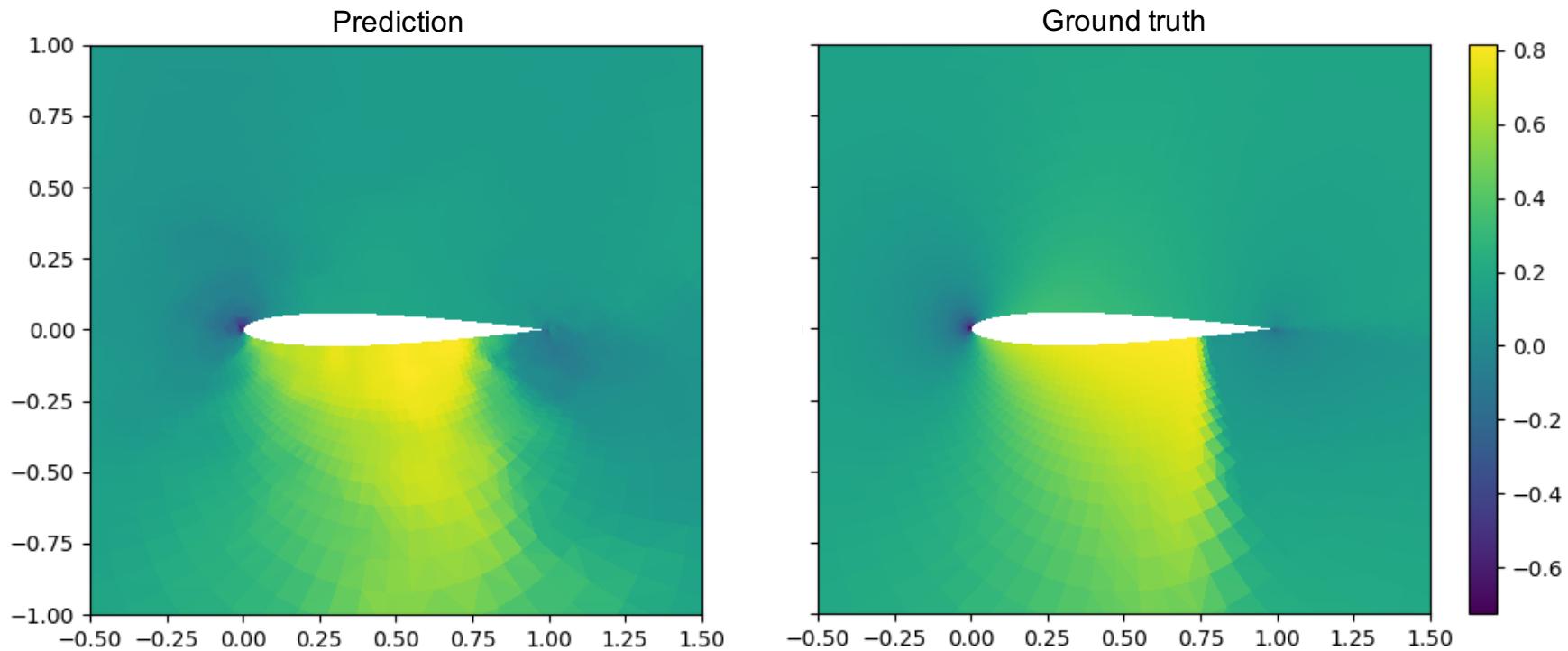
Generalization Results

- GCN Baseline



Generalization Results

- CFD-GCN



Timing Results

Model	Run Time
CFD-GCN	3s
Fine Mesh Simulation	189s

Summary

- By combining a coarse CFD simulation and GCN layers, the CFD-GCN model
 - Operates directly on meshes
 - Runs faster than a full CFD simulation
 - Achieves better generalization than a pure GCN model

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Thank you!