

# **An active-selecting neural network method for airfoil self-noise predictions**

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# Theory

- [Lighthill, 1952] → Conservation equations

$$\left( \frac{\partial^2}{\partial t^2} - c^2 \nabla^2 \right) \rho = \frac{\partial^2 T_{ij}}{\partial x_i \partial x_j} (\mathbf{x}, t) \quad (1)$$

Where  $T_{ij} = \rho u_i u_j + (p - c^2) \delta_{ij} - \tau_{ij}$  → Lighthill's stress tensor

- [Bogey et al., 2001] proposed the solution of acoustic pressure from Eq. 1:

$$p'(\mathbf{x}, t) = \frac{1}{4\pi c^2} \int_{V_y} \frac{r_i r_j}{r^3} \frac{\partial^2 T_{ij}}{\partial t^2} \left( \mathbf{y}, t - \frac{r}{c} \right) d\mathbf{y} \quad (2)$$

The acoustic pressure  $p'$  follows the dilatation field

(Large Eddy Simulation & hybrid method on Linearized Euler Equation):

$$\Theta = \nabla \cdot \mathbf{u} = -\frac{1}{\rho_0 c^2} \left( \frac{\partial p'}{\partial t} + U_i \frac{\partial p'}{\partial x_i} \right) \quad (3)$$

# Theory

- With the novel methods [Casalino, 2003] & [Najafi-yazdi et al., 2010] applying control surface function  $f(\mathbf{x}, t) = 0$ , we can further derive the acoustic pressure at any observer from Eq. 2:

$$p'(\mathbf{x}, t) = \frac{\partial}{\partial t} \int_{f=0} \left( \frac{Q_i n_i}{4\pi |\mathbf{x} - \mathbf{y}|} \right)_{t_e} dS - \frac{\partial}{\partial x_i} \int_{f=0} \left( \frac{L_{ij} n_j}{4\pi |\mathbf{x} - \mathbf{y}|} \right)_{t_e} dS + \frac{\partial^2}{\partial x_i \partial x_j} \int_{f>0} \left( \frac{T_{ij}}{4\pi |\mathbf{x} - \mathbf{y}|} \right)_{t_e} dV \quad (4)$$

Where

$$\left. \begin{aligned} Q_i &= \rho(u_i - v_i) + \rho_0 v_i \\ L_{ij} &= \rho u_i (u_j - v_j) + P_{ij} \end{aligned} \right\} \text{Source terms} \quad (5)$$

$$T_{ij} = \rho u_i u_j + ((p - p_0) - c^2(\rho - \rho_0)) \delta_{ij} - \tau_{ij} \longrightarrow \text{Lighthill's stress tensor}$$

$$P_{ij} = (p - p_0) \delta_{ij} - \tau_{ij} \longrightarrow \text{Compression tensor}$$

# Theory

- Recall the definition of sound pressure level SPL:

$$SPL = -20 \log \left( \frac{p'}{p_0} \right), \text{ in dB} \quad (6)$$

- Substitute Eq. 2 into Eq. 6, we have:

$$\begin{aligned} &SPL \\ &= -20 \left( \log \left( \frac{\partial}{\partial t} \int_{f=0} \left( \frac{\rho(u_i - v_i)n_i + \rho_0 v_i n_i}{4\pi |\mathbf{x} - \mathbf{y}|} \right)_{t_e} dS \right. \right. \\ &\quad \left. \left. - \frac{\partial}{\partial x_i} \int_{f=0} \left( \frac{(\rho u_i (u_j - v_j) + (p - p_0)\delta_{ij} - \tau_{ij})n_j}{4\pi |\mathbf{x} - \mathbf{y}|} \right)_{t_e} dS \right. \right. \\ &\quad \left. \left. + \frac{\partial^2}{\partial x_i \partial x_j} \int_{f>0} \left( \frac{(\rho u_i u_j + ((p - p_0) - c^2(\rho - \rho_0))\delta_{ij} - \tau_{ij})}{4\pi |\mathbf{x} - \mathbf{y}|} \right)_{t_e} dV \right) - \log p_0 \right) \end{aligned} \quad (7)$$

# Theory

- From Eq. 7 & Eq. 2, we conclude:

$$SPL = SPL(x, u, p, \rho, t, r, \mu) \quad (8)$$

- Recall the scaled  $SPL_{1/3}$  calculation (1/3 Octave band) [Brooks et al., 1989]:

$$\text{Scaled } SPL_{1/3} = SPL_{1/3} - 10\log\left(M^5 \frac{\delta\Lambda}{r^2}\right) \bar{D} \quad (9)$$

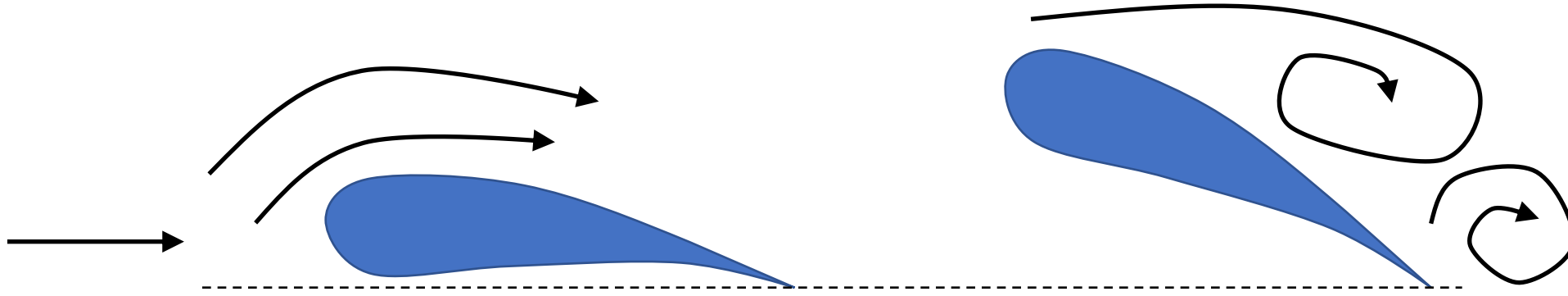
$$\longrightarrow \text{Scaled } SPL_{1/3} = \text{Scaled } SPL_{1/3}(x, u, p, \rho, t, r, \mu, \Lambda) \quad (10)$$

- Recall the NASA database [Brooks et al., 1989], there are five input attributes:

$$\text{Scaled } SPL_{1/3} = \text{Scaled } SPL_{1/3}(\alpha, f, l, u, \delta) \quad (11)$$

Where  $\delta = \delta(\mu)$ .

# Simulation



Flow field:  $\{\rho, u, p\} \rightarrow \alpha = \alpha(\rho, u, p)$

- Frequency  $f$  corresponds with the  $SPL_{1/3}$  from the simulation as output.
- Chord length  $l$  is scaled to be 1 in the simulation.
- Hence, the experimental data can be recognized as

$$\text{Scaled } SPL_{1/3} = \text{Scaled } SPL_{1/3}(\alpha(\rho, u, p), f, l, u, \delta(\mu)) \quad (12)$$

$$(\text{Scaled } SPL_{1/3}, f) = (\text{Scaled } SPL_{1/3}, f)(\rho, u, p, \mu) \quad (13)$$

# Simulation

- Compare Eq. 8 with Eq. 12:

$$SPL_{1/3} = SPL_{1/3}(x, u, p, \rho, t, r, \mu)$$

$$(\text{Scaled } SPL_{1/3}, f) = (\text{Scaled } SPL_{1/3}, f)(\rho, u, p, \mu)$$

- In the actual simulation, the flow field  $\{\rho, u, p\}$  &  $(\text{Scaled } SPL_{1/3}, f)$  is the output data;  
 $\{\mu, \text{Turbulence Model}, U, \alpha\}$ , where  $\text{Turbulence Model} = \text{Turbulence Model}(k, \varepsilon, C_\mu, \mu, l)$  for k- $\varepsilon$ .
- Therefore, the simulation model can be simplified as:

$$(\rho, u, p, (\text{Scaled } SPL_{1/3}, f)) = (\rho, u, p, (\text{Scaled } SPL_{1/3}, f))(\mu, \text{Turbulence Model}, U, \alpha) \quad (14)$$

- Recall the neural network on the experimental data:

$$\text{Scaled } SPL_{1/3} = NN_{\text{exp}}(\alpha, f, l, U, \delta) \quad (15)$$

# Turbulence Model

- **K-ε model:**
- Turbulent kinetic energy  $k$  and dissipation  $\varepsilon$  obeys

$$\frac{\partial(\rho k)}{\partial t} + \frac{\partial(\rho k u_i)}{\partial x_i} = \frac{\partial}{\partial x_i} \left( \frac{\mu_t}{\sigma_k} \frac{\partial k}{\partial x_j} \right) + 2\mu_t E_{ij} E_{ij} - \rho \varepsilon \quad (16)$$

$$\frac{\partial(\rho \varepsilon)}{\partial t} + \frac{\partial(\rho \varepsilon u_i)}{\partial x_i} = \frac{\partial}{\partial x_i} \left( \frac{\mu_t}{\sigma_\varepsilon} \frac{\partial \varepsilon}{\partial x_j} \right) + C_{1\varepsilon} \frac{\varepsilon}{k} 2\mu_t E_{ij} E_{ij} - C_{2\varepsilon} \rho \frac{\varepsilon^2}{k} \quad (17)$$

Where  $u_i$ : velocity components;  $E_{ij}$ : rate of deformation;  $\mu_t \rightarrow$  eddy viscosity  $\frac{\mu}{\mu_t} = \frac{\rho C_\mu k^2}{\mu \varepsilon}$

- The terms  $k$  and  $\varepsilon$  is calculated as:

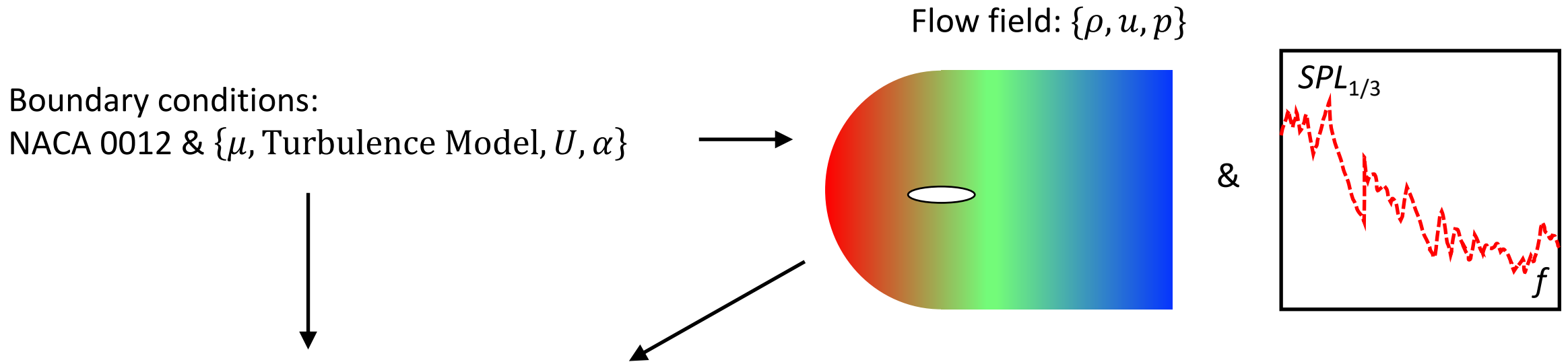
$$k = \frac{3}{2} (Ul)^2 \text{ \& } \varepsilon = \frac{C_\mu^{\frac{3}{4}} k^{\frac{3}{2}}}{l'} \quad (l' \approx 0.7l) \quad (18)$$

The constants adopts:  $C_\mu = 0.09$ ,  $\sigma_k = 1.00$ ,  $\sigma_\varepsilon = 1.30$ ,  $C_{1\varepsilon} = 1.44$ ,  $C_{2\varepsilon} = 1.92$



# Neural Network

- With the  $NN_{\text{exp}}$ , we could fit the regression on self-noise experimental model;
- We can also apply a neural network on the simulation data corresponding the theoretical model (Eq. 13);
- The simulation process can be shown as:



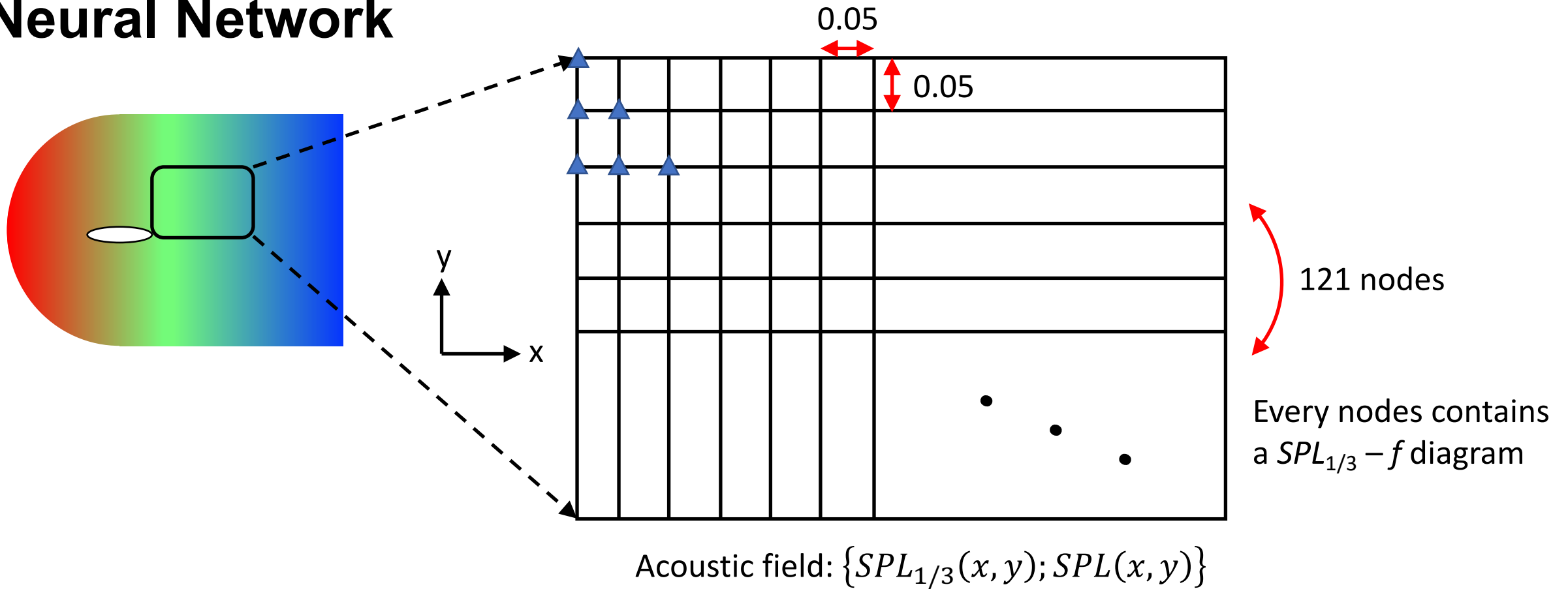
Input data:

Output data:

$\{\mu, \text{Turbulence Model}, U, \alpha, \text{Flow Field}(\rho, u, p)\} \rightarrow \text{Noise field: } SPL_{1/3} \text{ distribution;}$

$$SPL_{1/3}(x, y) = NN_{\text{sim}}\{\mu, \text{Turbulence Model}, U, \alpha, \text{Flow Field}(\rho, u, p)\}$$

# Neural Network



- Recall the simulation neural network:  $NN_{sim}\{\mu, \text{Turbulence Model}, U, \alpha, \text{Flow Field}(\rho, u, p)\}$ , we could further predict the acoustic field considering specific frequency.

# Neural Network

- With the set conditions:
- K-ε, FW-H model parameters:

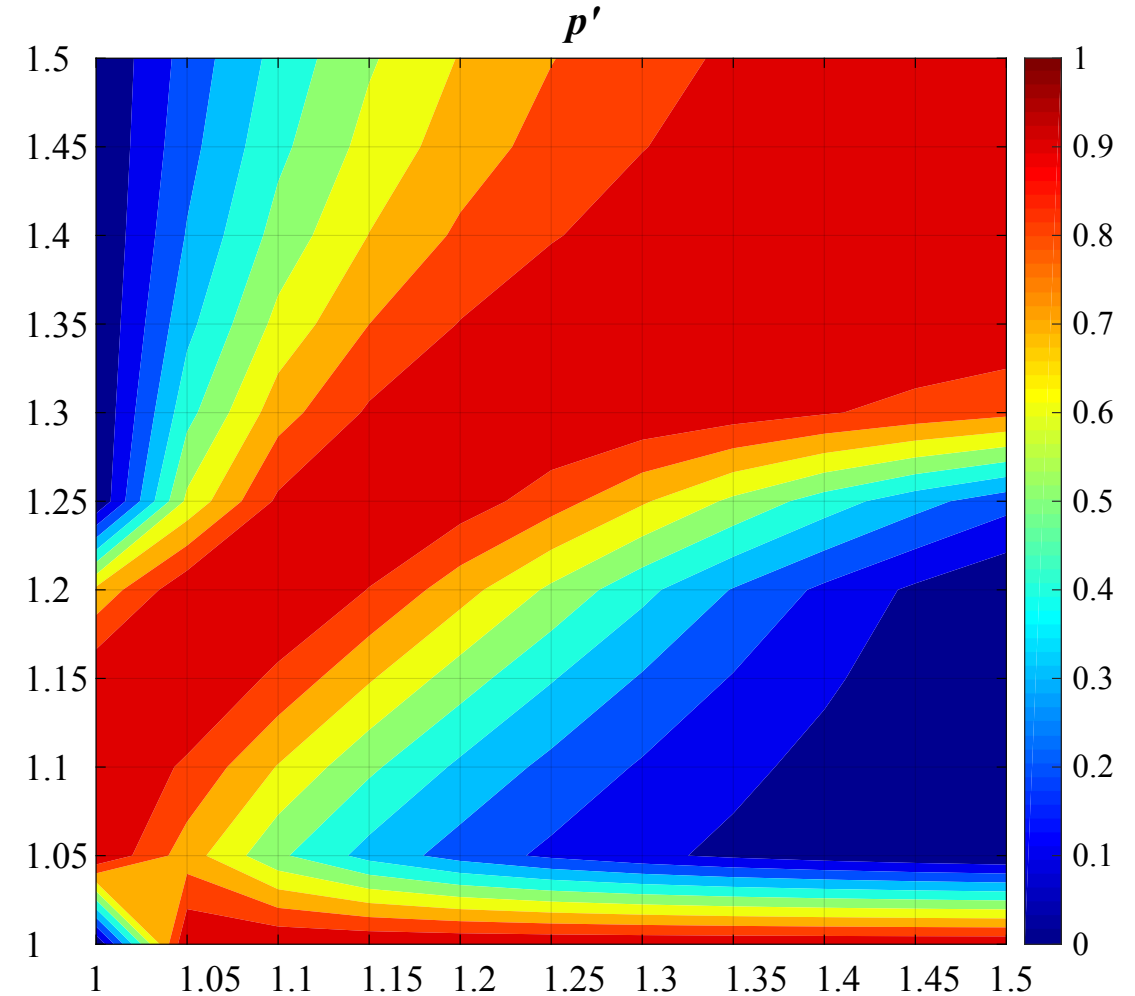
$$\left\{ \begin{array}{l} TKE Prandtl Number: 1 \\ TDR Prandtl Number: 1.3 \\ Energy Prandtl Number: 0.85 \\ Wall Prandtl Number: 0.85 \\ far field: \rho = 1.225 kg/m^3 \\ c = 340 m/s \end{array} \right.$$

- Boundary conditions:

$$\left\{ \begin{array}{l} far field: Mach Number = 0.2048 \\ k = 1 \text{ and } \varepsilon = 1 \\ T = 300K \\ roughness height = 0, roughness const = 0.5 \end{array} \right.$$

- From Eq. 2, acoustic pressure follows:

$$p'(\mathbf{x}, t) = \frac{1}{4\pi c^2} \int_{V_y} \frac{r_i r_j}{r^3} \frac{\partial^2 (\rho u_i u_j + (p - c^2) \delta_{ij} - \tau_{ij})}{\partial t^2} \left( \mathbf{y}, t - \frac{r}{c} \right) d\mathbf{y}$$

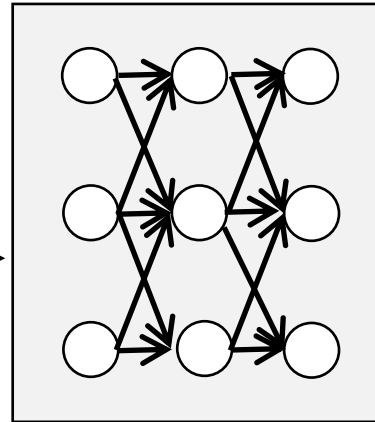
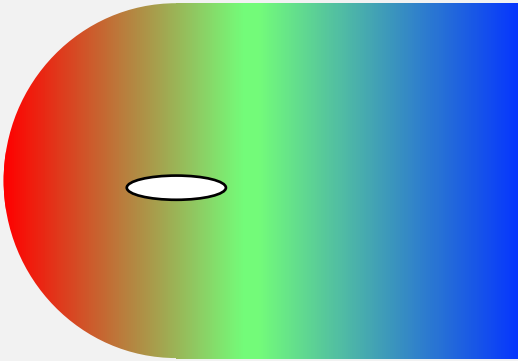


$$\rightarrow p' = p'(u, x, r, \mu, \rho, p, t).$$

# Simulation Coupled NN

## Field data:

- Pressure field  $p$
- Density field  $\rho$
- Velocity field  $u$
- Energy field  $E$
- Turbulence kinetic energy  $k$
- Turbulence intensity  $I_T$
- Turbulence dissipative rate  $\varepsilon$

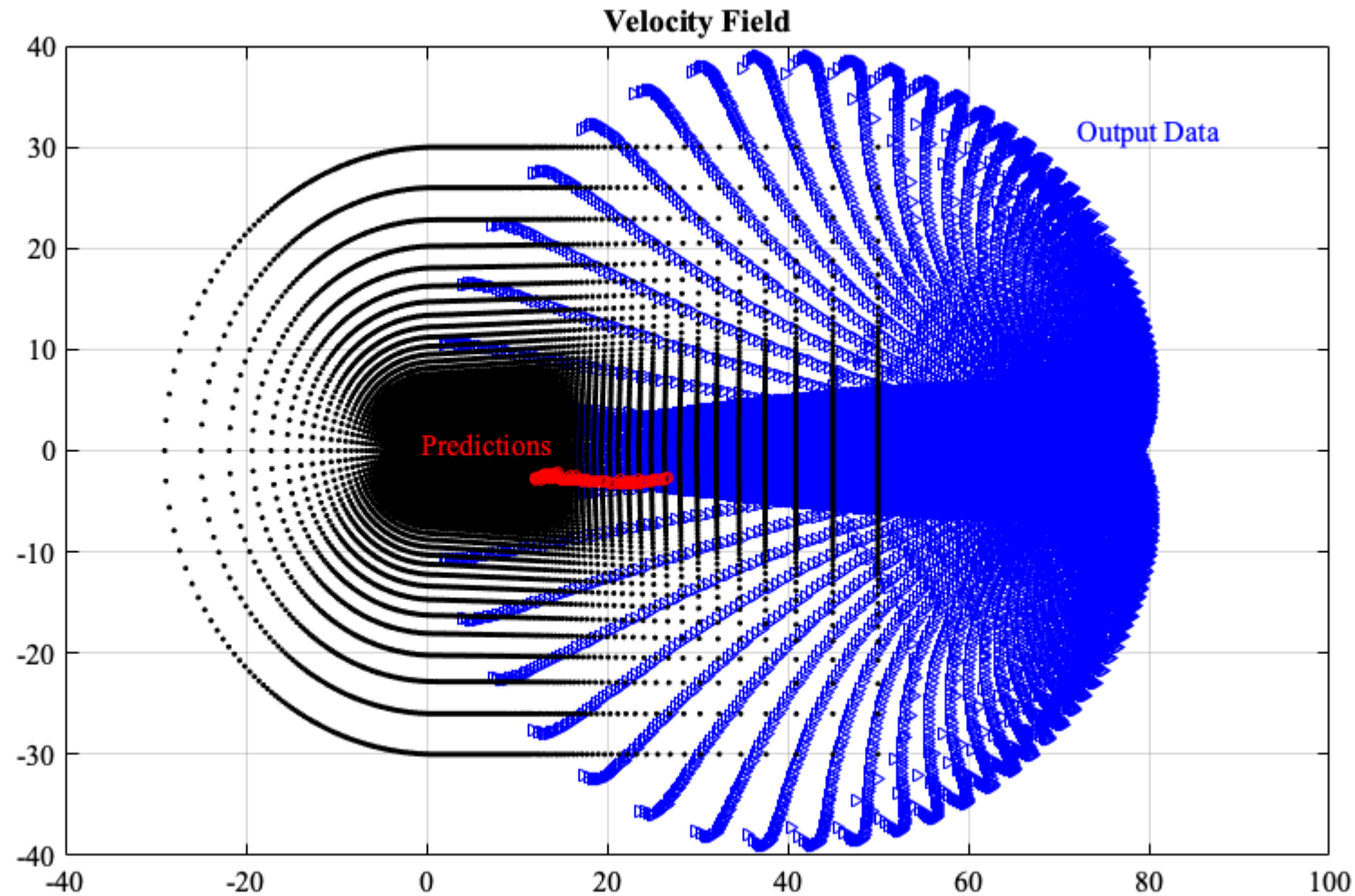


- Velocity field  $u$

# Simulation Coupled NN

***BAD Results!!!***

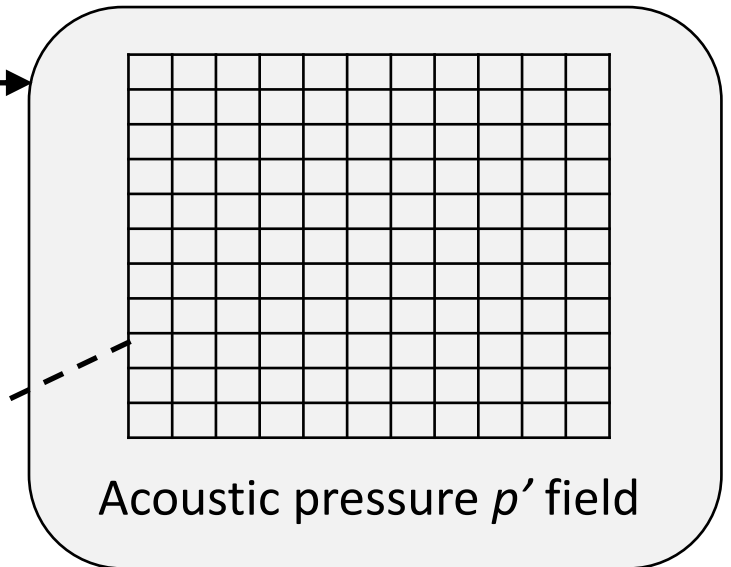
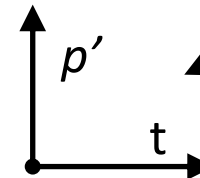
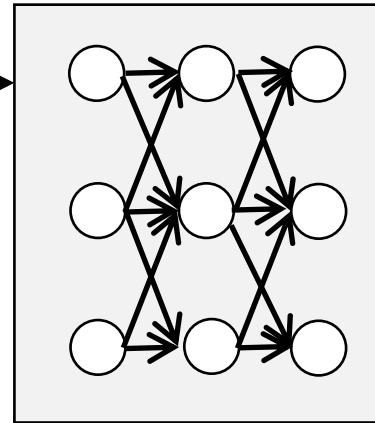
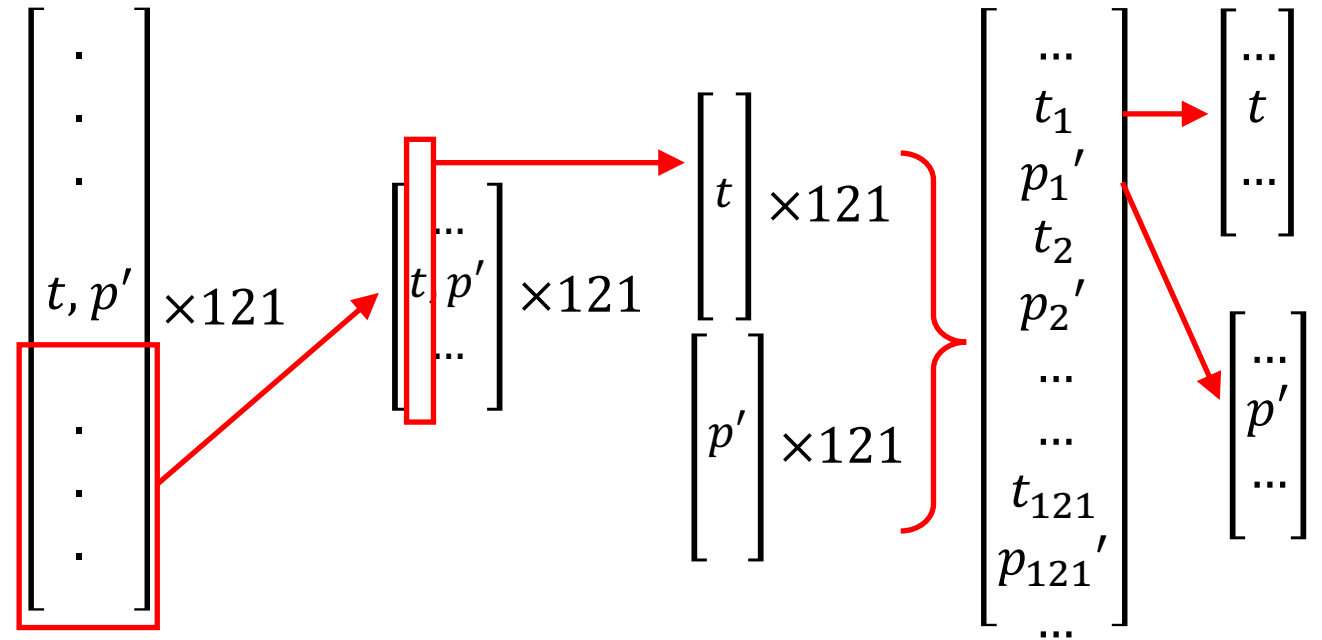
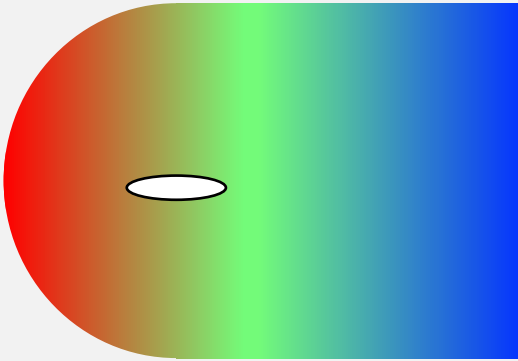
Because only simple case data is trained



# Simulation Coupled NN

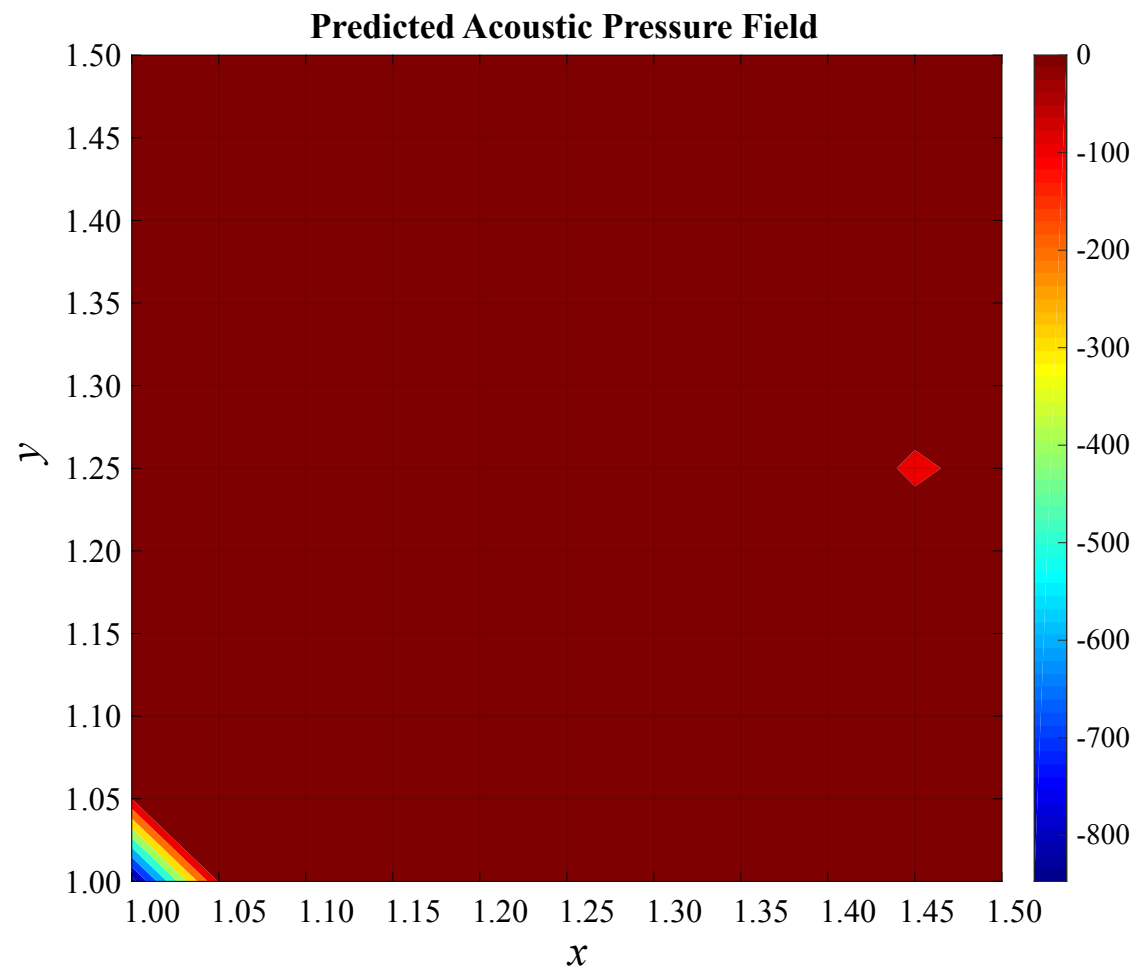
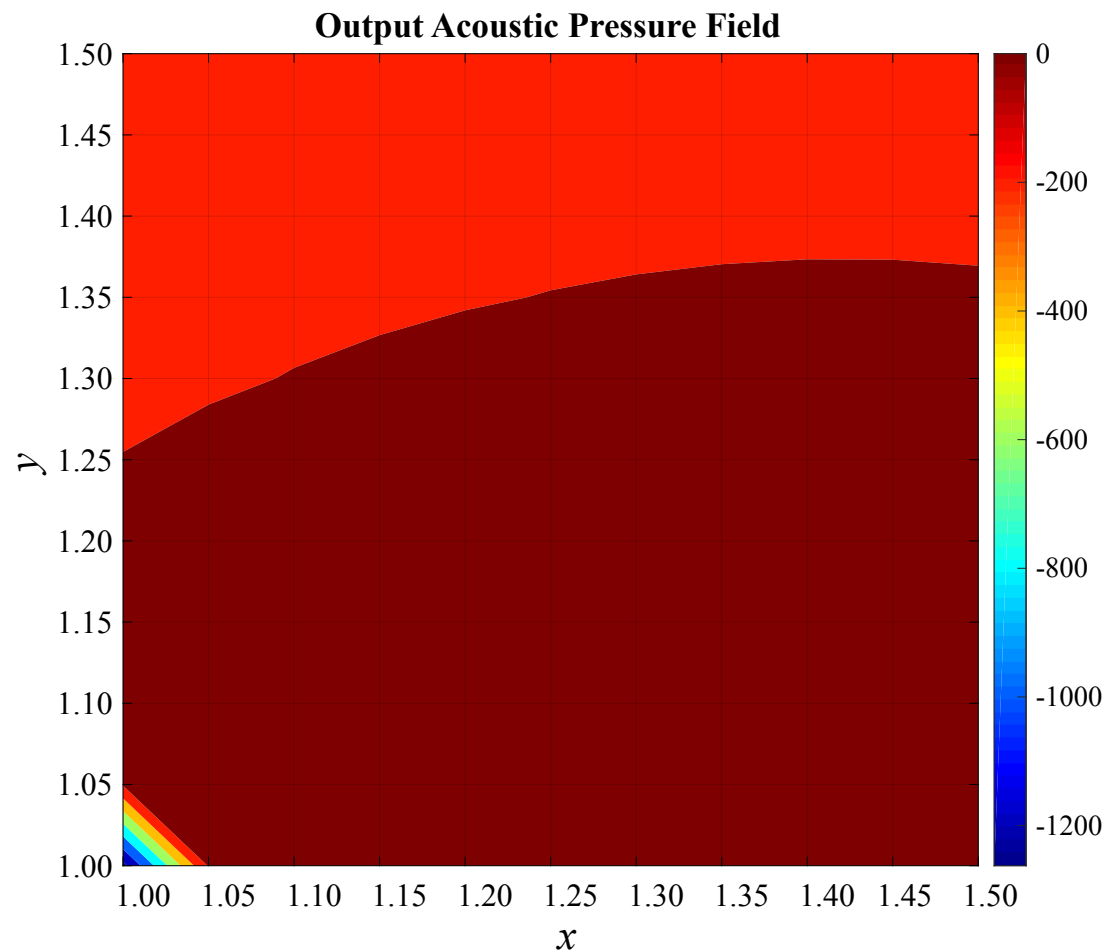
## Field data:

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- Turbulence dissipative rate  $\varepsilon$



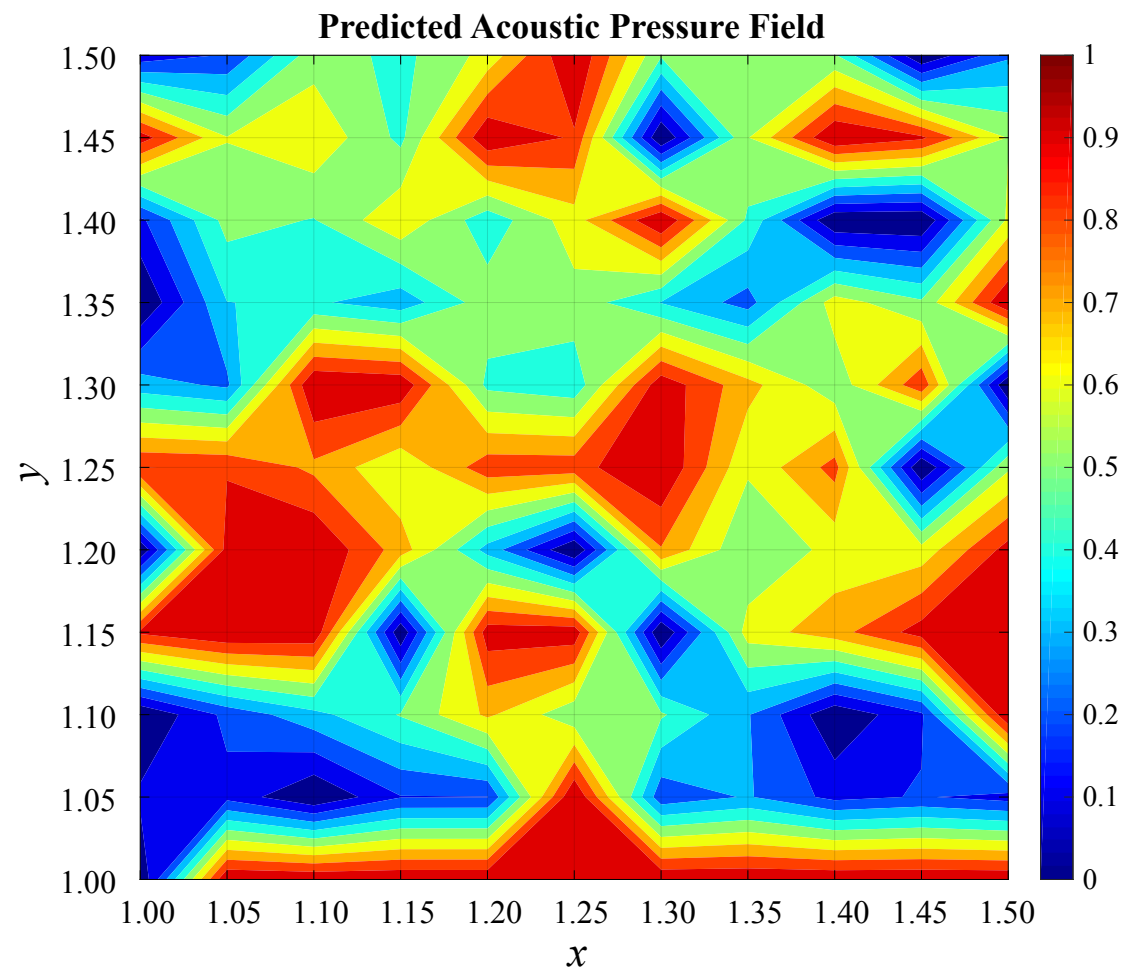
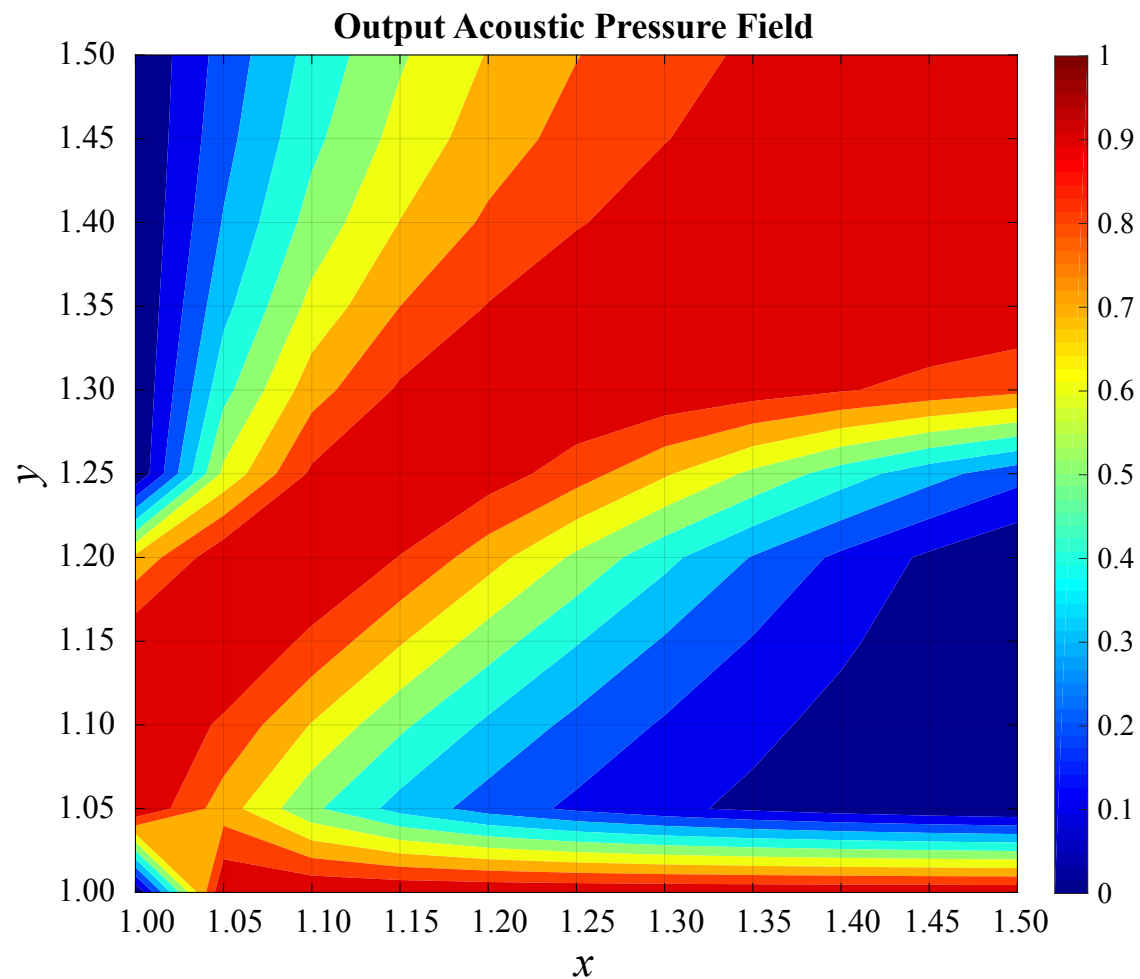
# Simulation Coupled NN

## Original Data at Final Time Step



# Simulation Coupled NN

## Normalized Data at Final Time Step

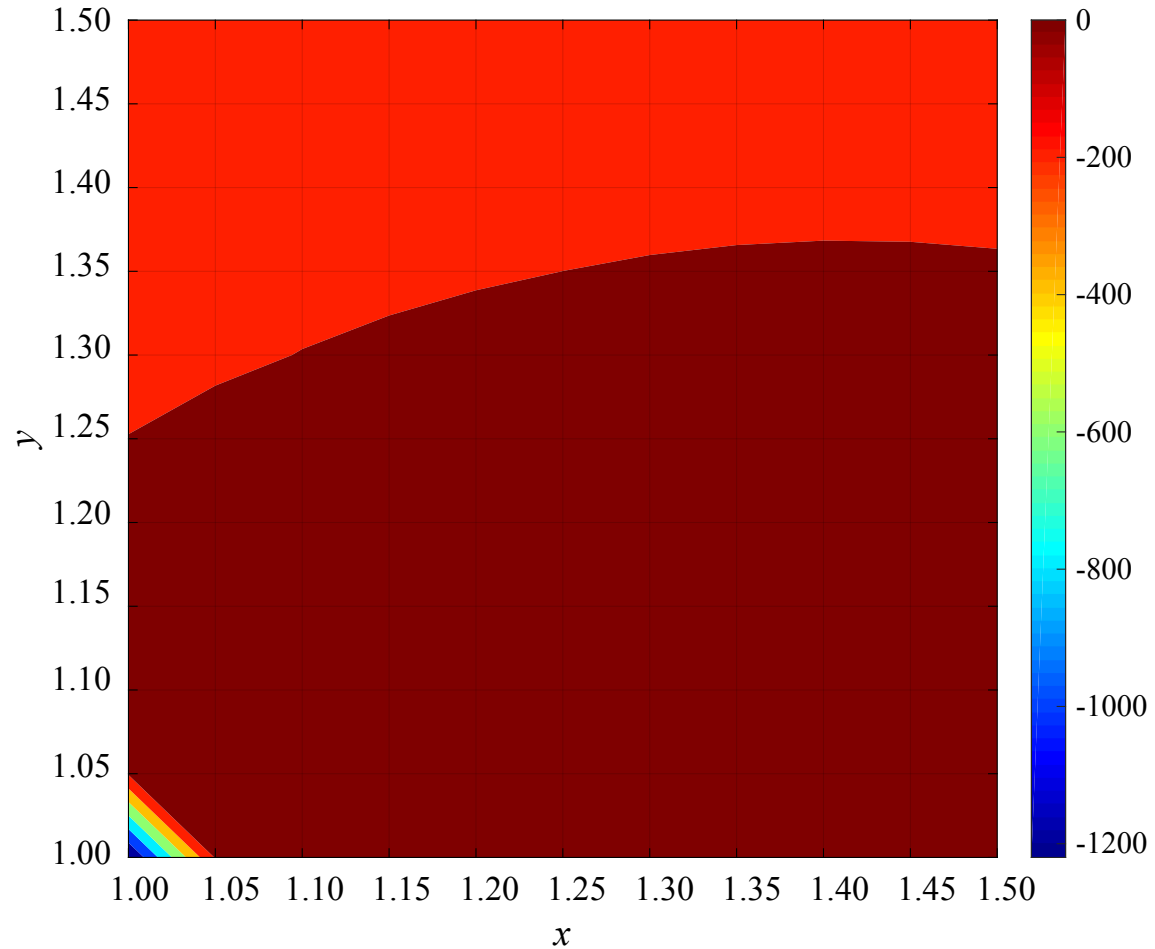




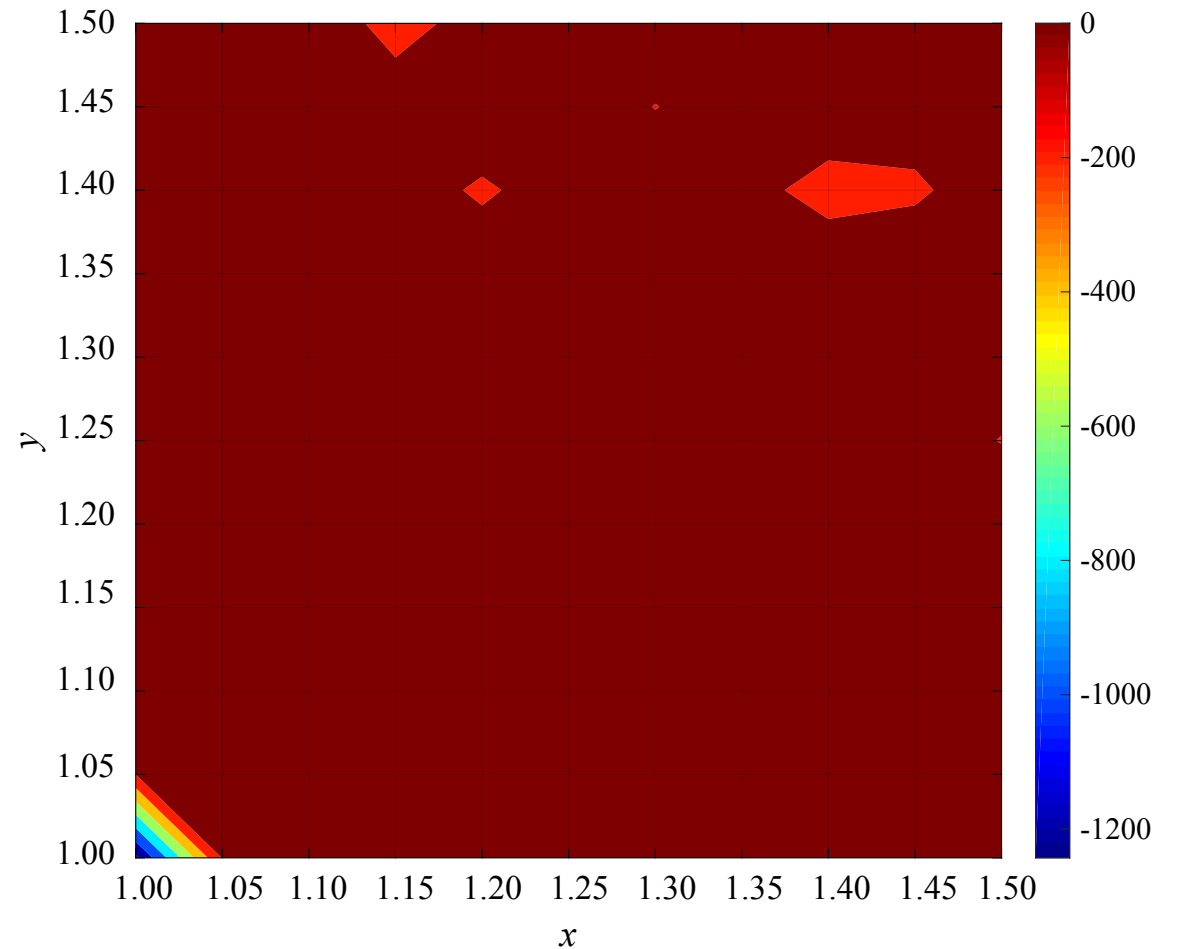
# Simulation Coupled NN

Original Data at Final Time Step - 600

CFD



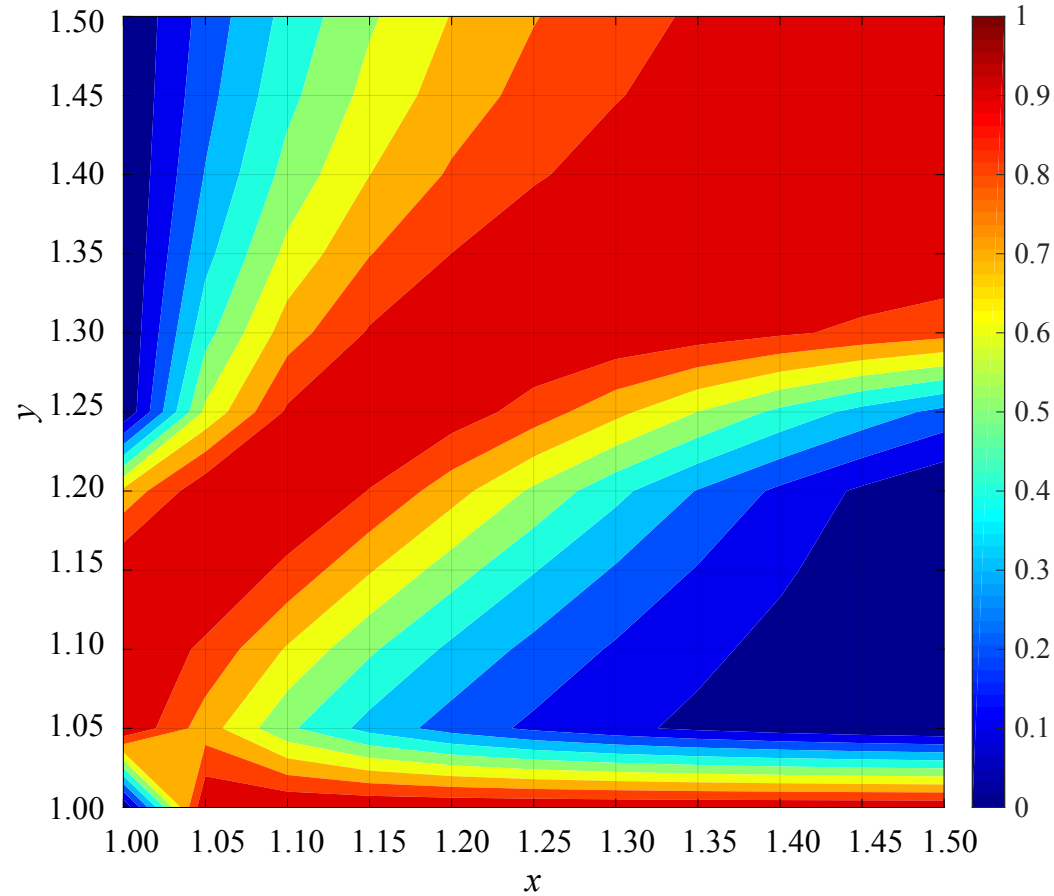
NN



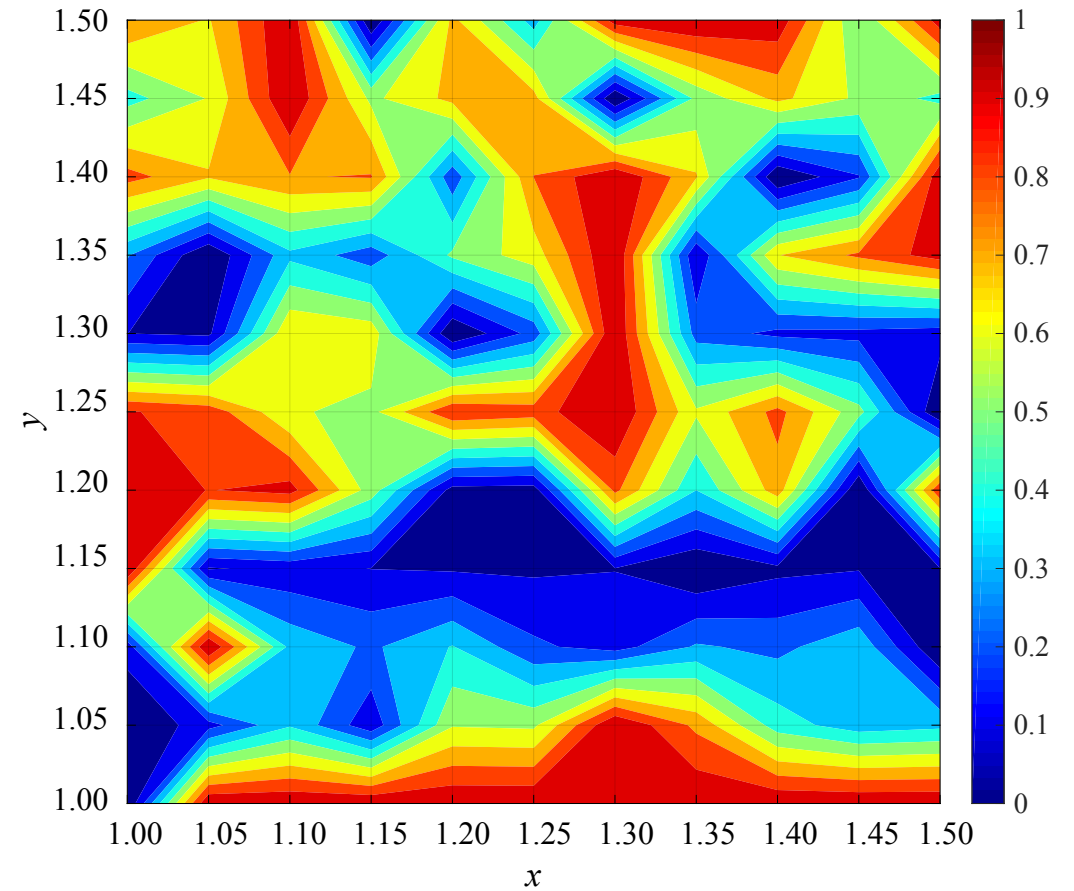
# Simulation Coupled NN

Normalized Data at Final Time Step - 600

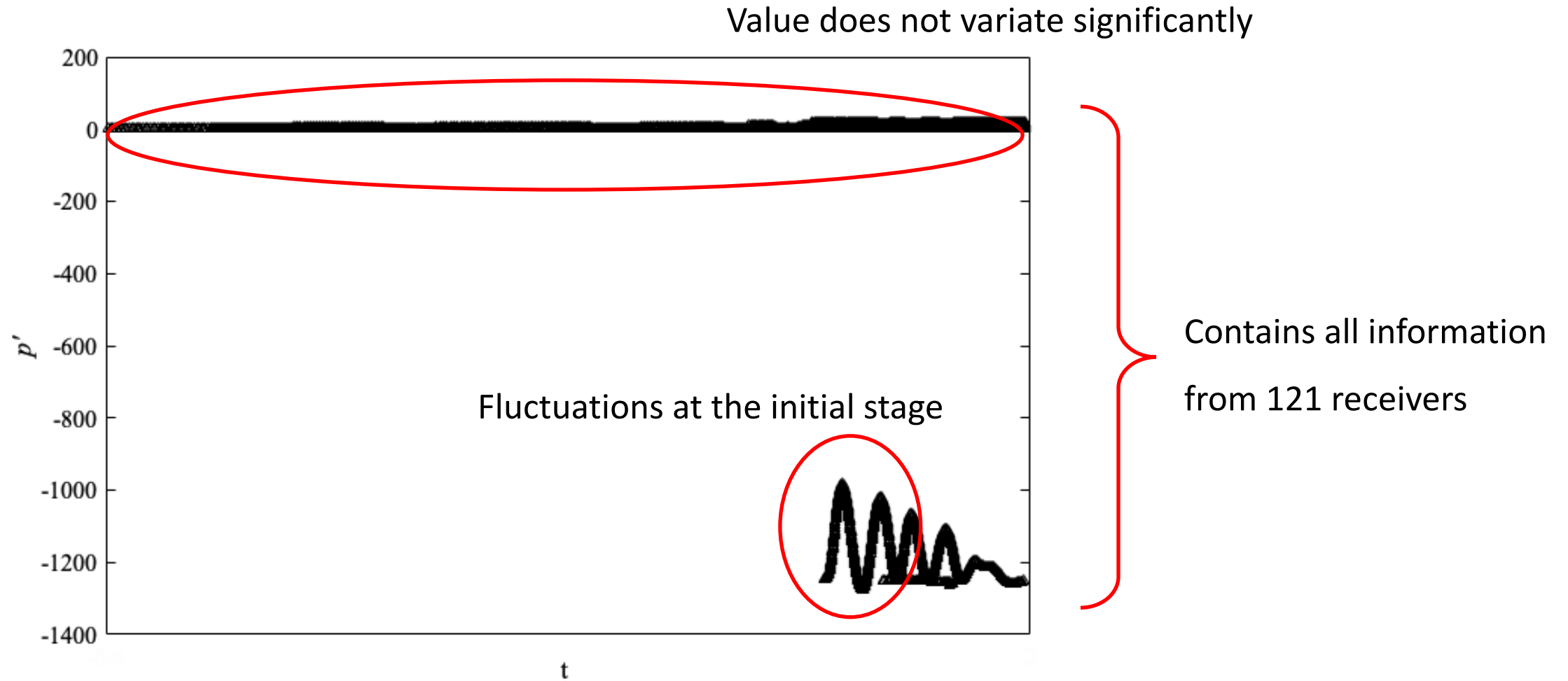
CFD



NN

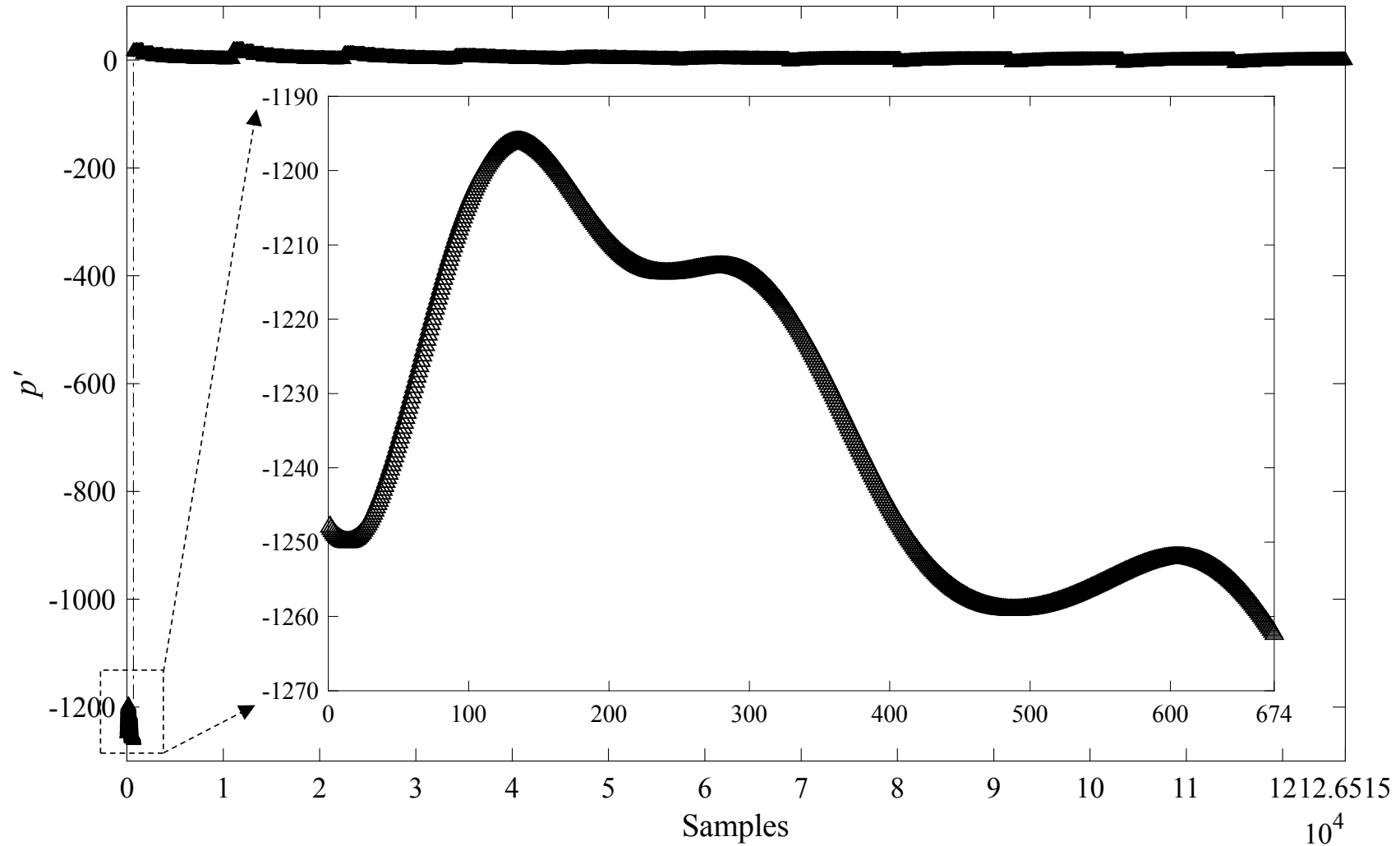


# Simulation Coupled NN



# Simulation Coupled NN

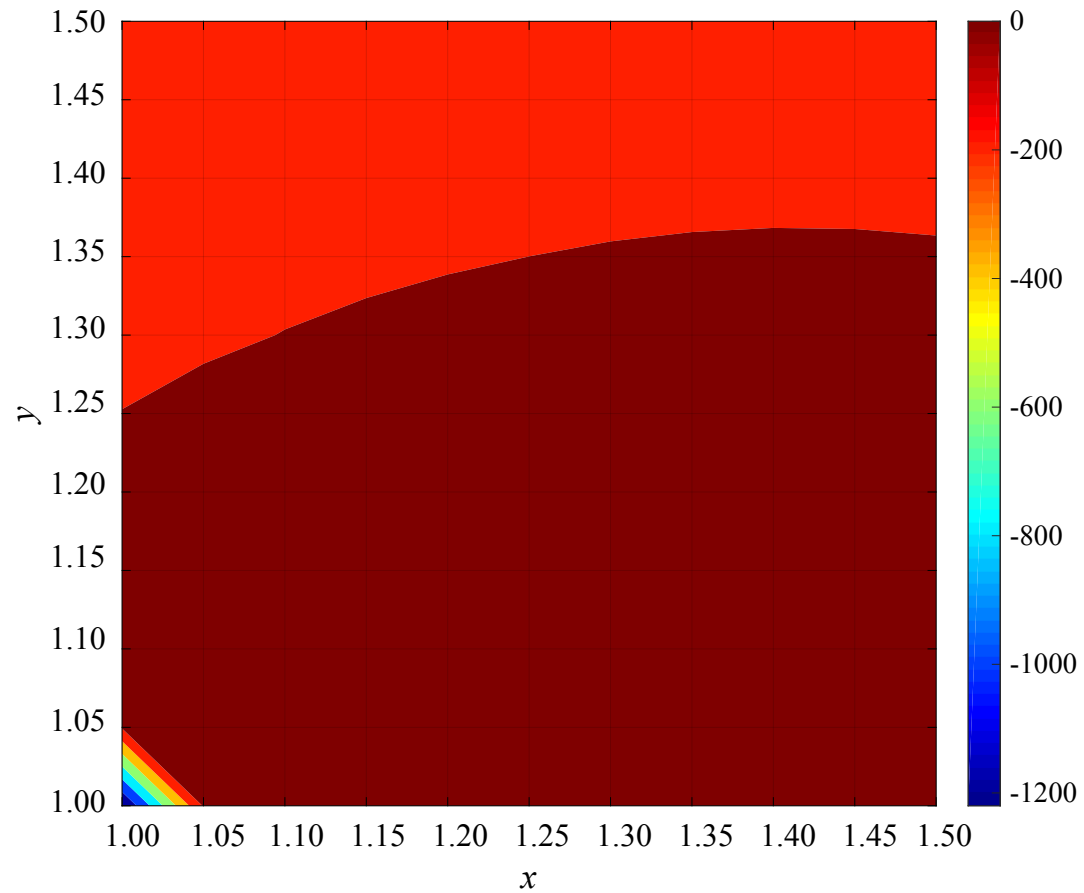
- *How to capture the detailed noise signals*



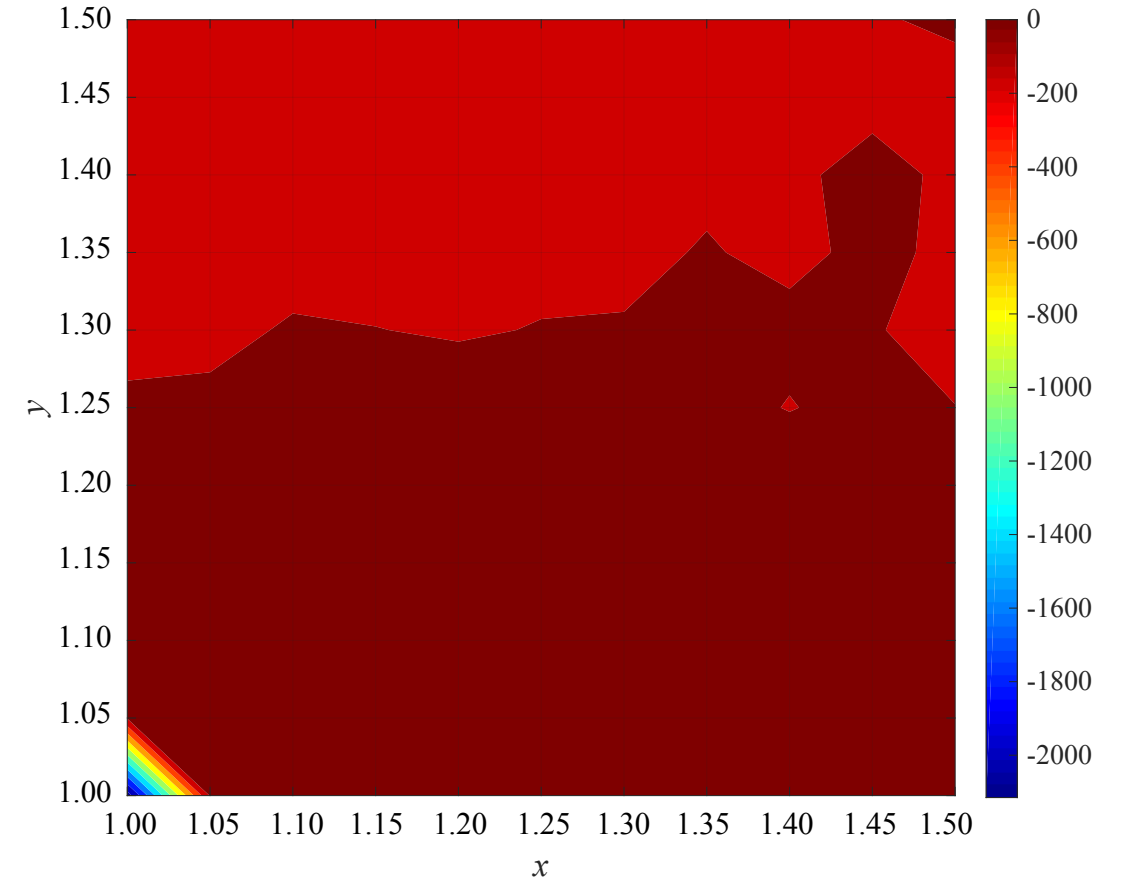
# Simulation Coupled NN

Original Data at Final Time Step

CFD



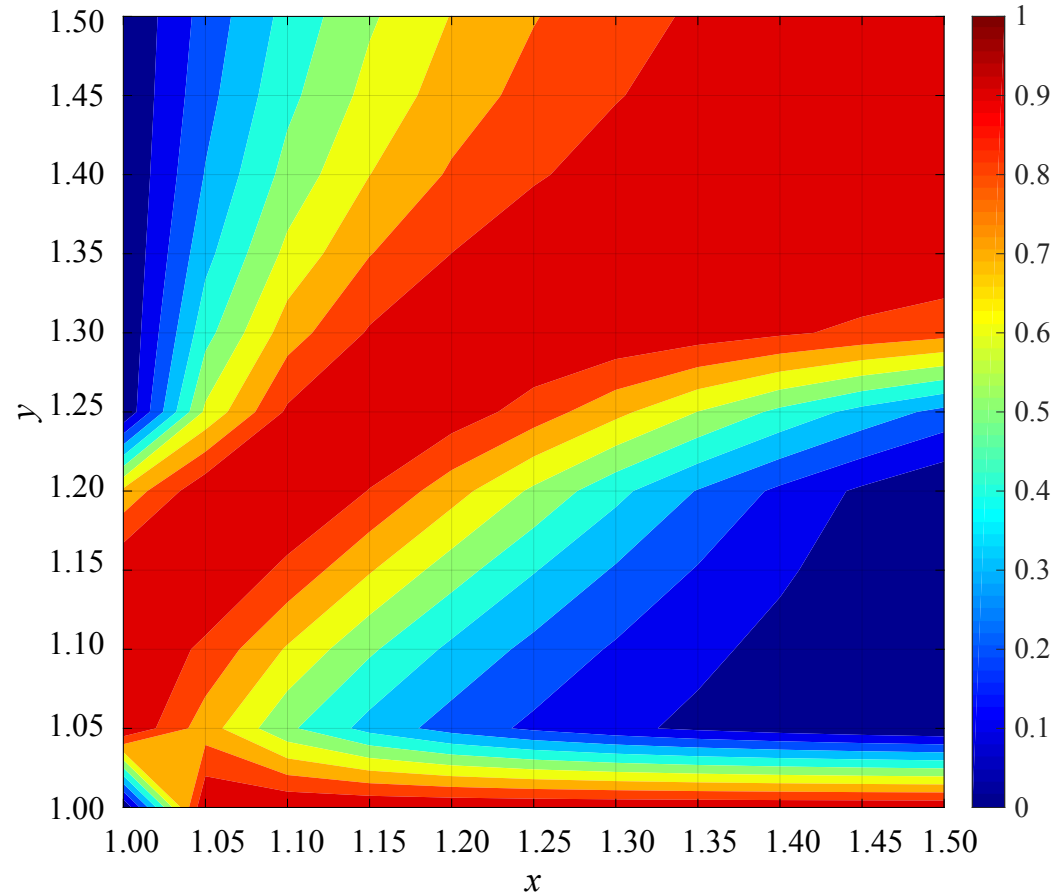
NN



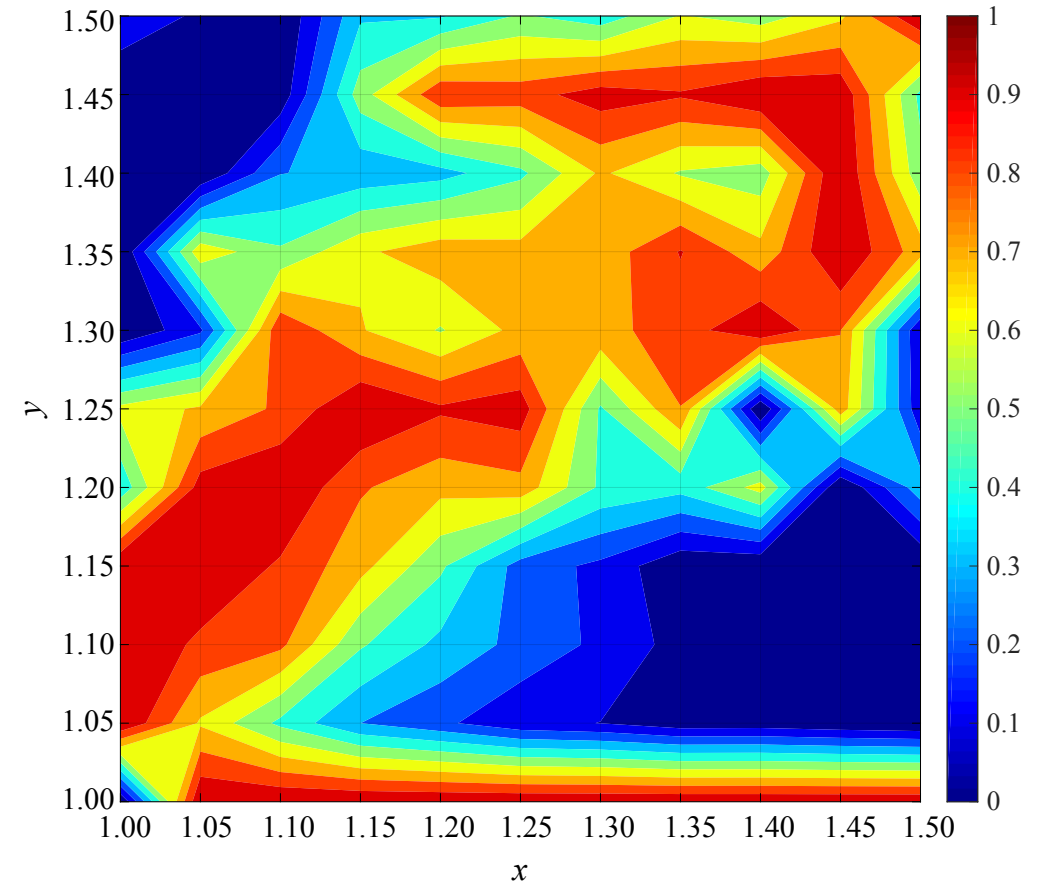
# Simulation Coupled NN

Normalized Data at Final Time Step

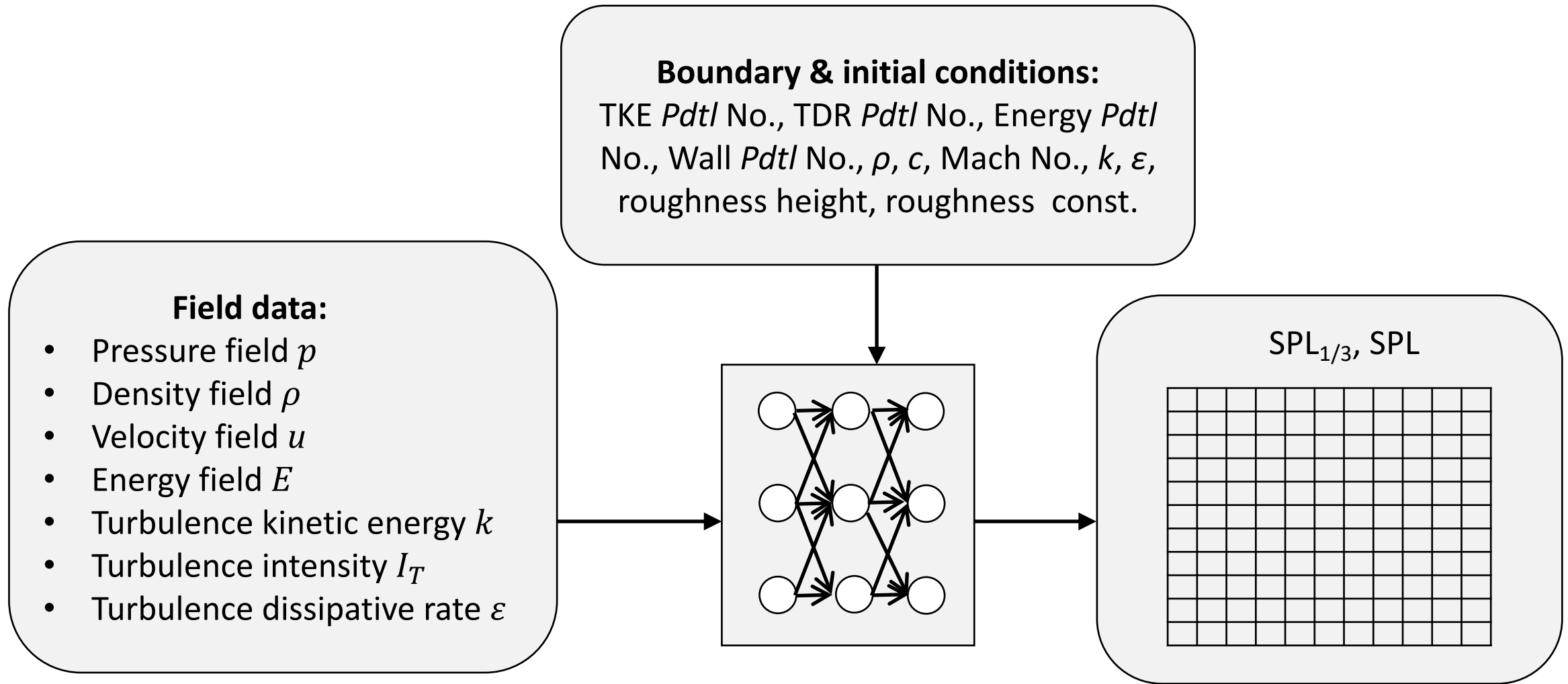
CFD



NN



# Simulation Coupled NN



# Main Problems

- How to adopt the predictions ( $NN_{\text{exp}}$ ) from the NASA database.
- How to accumulate the physics law into the  $NN_{\text{sim}}$  predictions.
- The chord length of airfoil.
- Adjust the simulation parameters to the input data.

*- The End -*