

Reconstruction of plasma equilibrium using physics-informed neural network on EAST

**Wenbin Liao^{1,2*}, Zhengping Luo^{1*}, Yao Huang^{1,3},
Yuehang Wang¹, Kai Wu¹, Zijie Liu³, Bingjia Xiao^{1,2,3}**

¹ Institute of Plasma Physics, Hefei Institutes of Physical Science, Chinese Academy of Sciences

² University of Science and Technology of China

³ Institute of Energy, Hefei Comprehensive National Science Center

*Email: wenbin.liao@ipp.ac.cn & zhpluo@ipp.ac.cn



Equilibrium Reconstruction – Solving Grad-Shafranov equation

G-S equation:
$$R \frac{\partial}{\partial R} \frac{1}{R} \frac{\partial \psi}{\partial R} + \frac{\partial^2 \psi}{\partial Z^2} = -\mu_0 R^2 \frac{d}{d\psi} p - F \frac{d}{d\psi} F$$

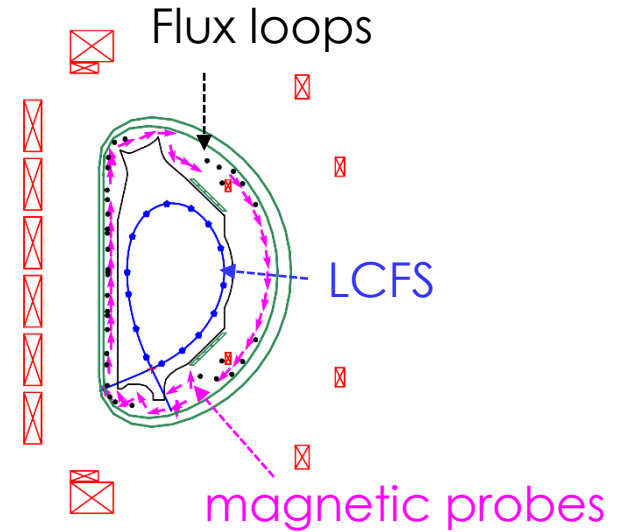
□ **Known:**

External diagnostics, e.g., plasma current I_p , magnetic probes, flux loops, etc.

□ **Solving:**

Internal diagnostics, e.g., **MSE**, **POINT**;

$\psi(R, Z)$ that satisfies the G-S equation on the domain Ω_c .



EFIT:

□ **Distributed current model:**

$$p'(\psi) = \sum_n \alpha_n \psi^n \quad (\text{H-mode}) \quad \text{-----} \rightarrow \text{Difficult for complex form}$$

$$F(\psi)F'(\psi) = \sum_n \gamma_n \psi^n$$

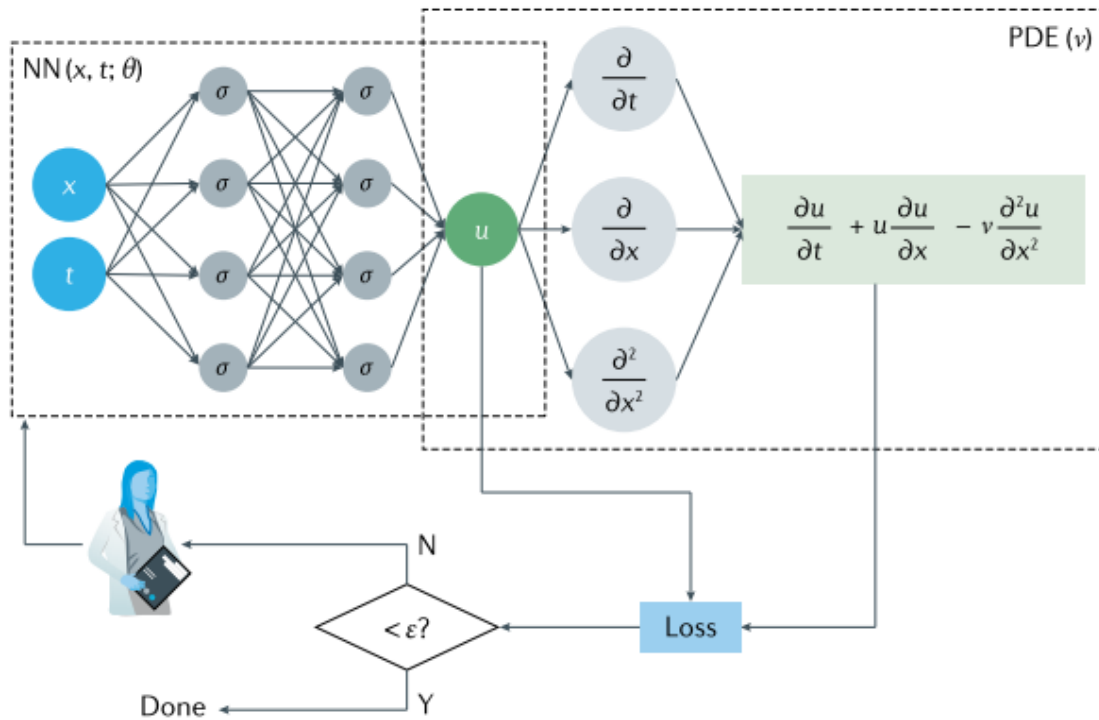
□ **Picard iterations:**

$$C_j^{(m+1)}(r_j, z_j) = \sum_{n=1}^{n_{PF}} G_{CD_j}(r_j, z_j, r_n, z_n) I_n^{(m+1)} + \int_{\Omega^{(m)}} dR dZ G_{pD_j}(r_j, z_j, R', Z') J_\phi(R', Z', \psi^{(m)}, \vec{\alpha}^{(m+1)}, \vec{\gamma}^{(m+1)})$$

□ **Least squares fitting:**

$$\chi^2 = \sum_{n=1}^{n_M} \left(\frac{M_j - C_j}{\sigma_j} \right)^2 \quad \text{-----} \rightarrow \text{Difficult to add new diagnostics}$$

Physics-Informed Neural Network (PINN)

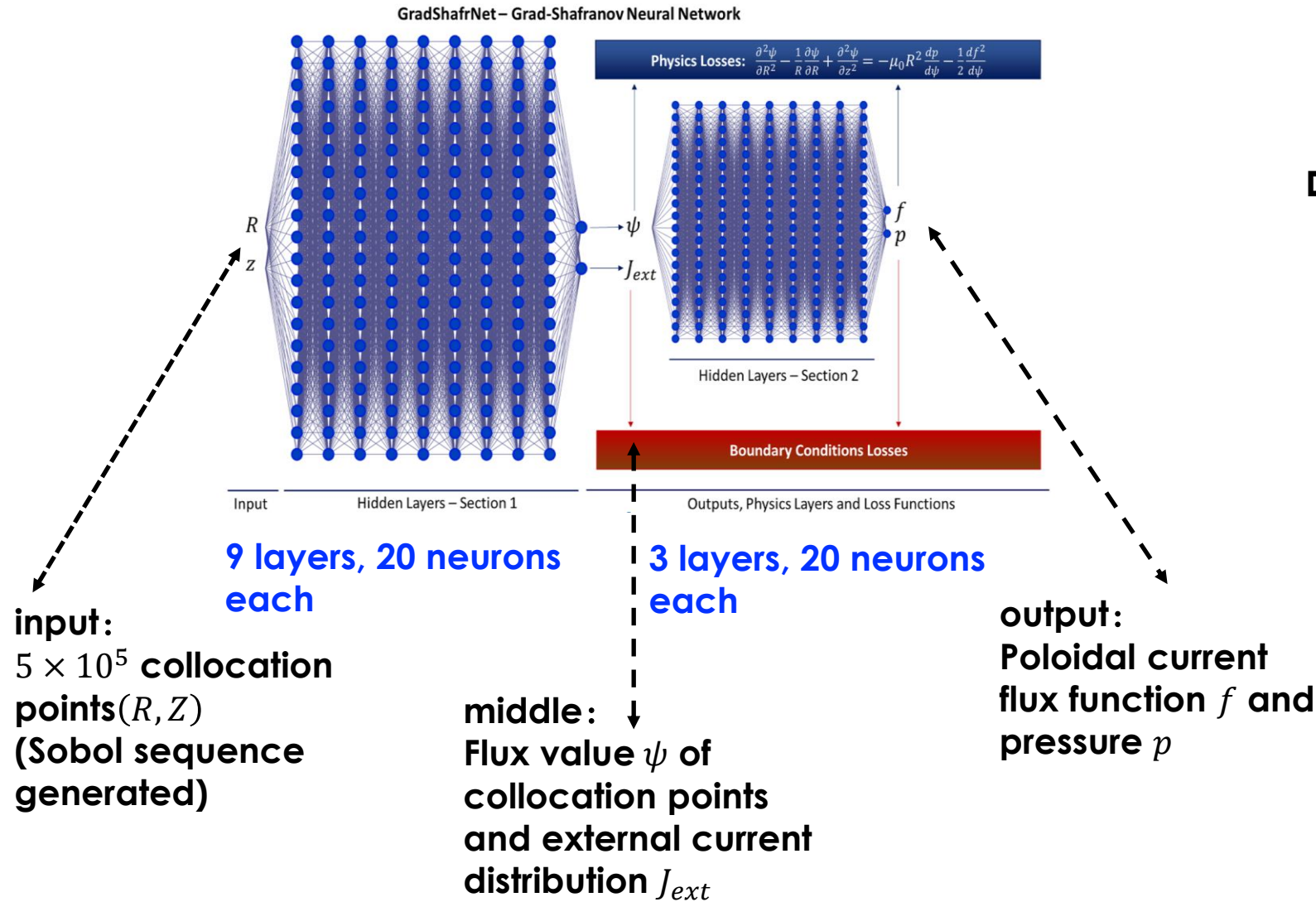


The basic structure of PINN

	Traditional numerical methods	PINN
Model building	$p'(\psi) = \sum_n \alpha_n \psi_N^n$ $F(\psi)F'(\psi) = \sum_n \gamma_n \psi_N^n$	$p'(\psi) = NN_1(\psi)$ $F(\psi)F'(\psi) = NN_2(\psi)$
New diagnostics	New response matrix, new model assumptions, etc.	New boundary conditions
Applicability to complex problems	Difficult	Easy

- PINN can be used for plasma equilibrium reconstruction.

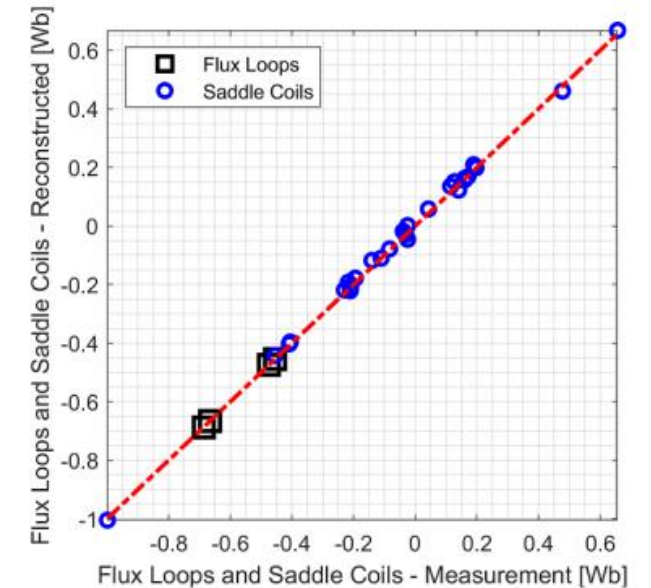
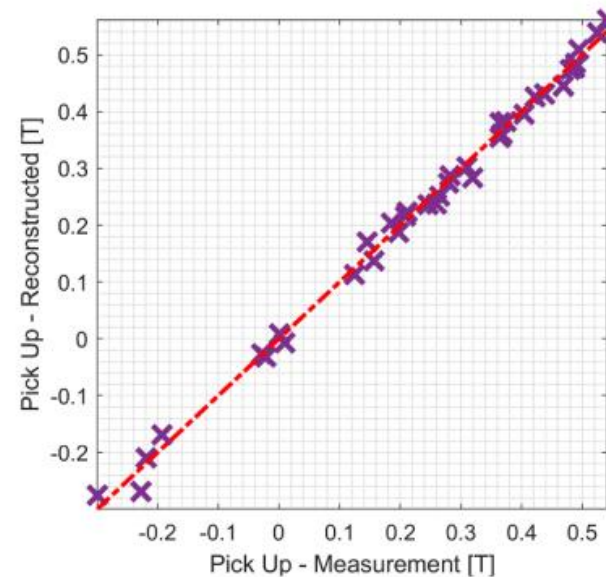
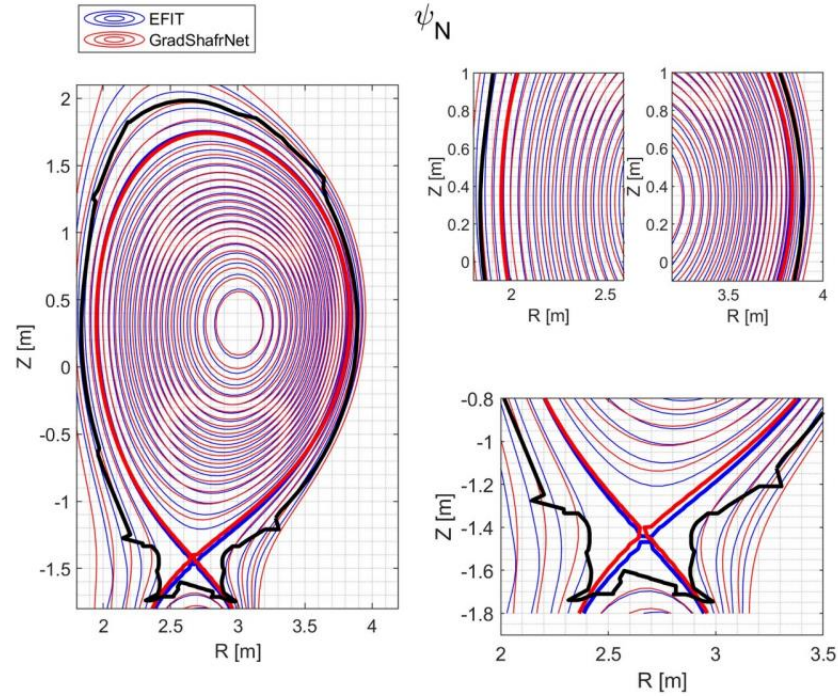
Reconstruction by PINN on JET



□ Optimizer: Adam

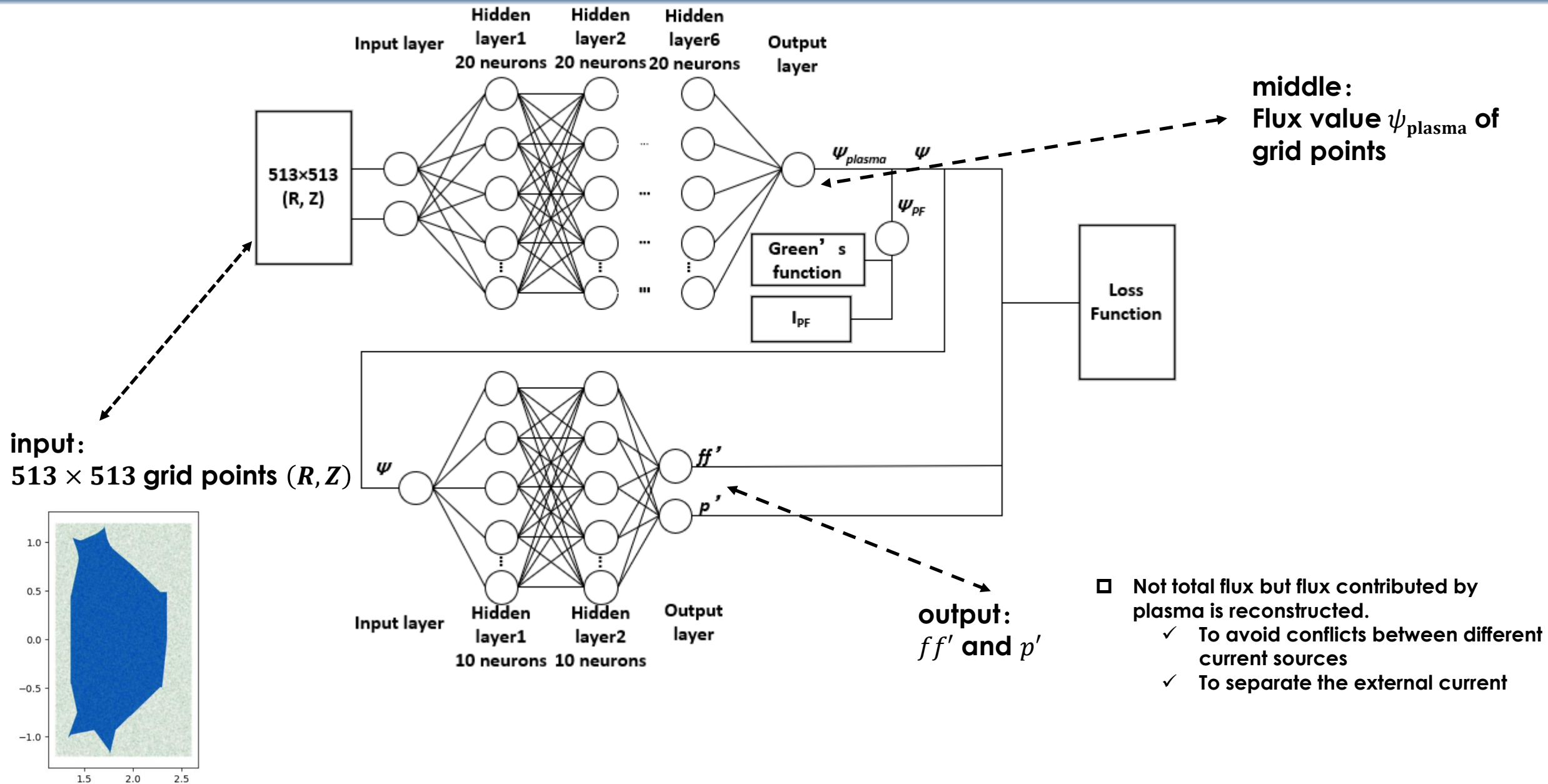
- Boundary Conditions Losses:
- Pick-up coils
- Flux loops
- Saddle coils
- Pressure on the wall

□ Shot 94217 t=9s



- ✓ Relative error of $\psi_N < 5\%$
- ✓ Measurements and predictions in good agreement
- Unreasonable assumptions of external current distribution
- Conflation of plasma boundary and limiter
- Great error

PINN Structure for reconstruction on EAST (I_{pF} known)



Loss Functions (I_{pF} known)

$$\mathcal{L} = w_{p,inside} \mathcal{L}_{p,inside} + w_{p,outside} \mathcal{L}_{p,outside} + w_{b,MP} \mathcal{L}_{b,MP} + w_{b,FL} \mathcal{L}_{b,FL} + w_{p',p,LCFS} \mathcal{L}_{p',p,LCFS} + w_{ff',p,LCFS} \mathcal{L}_{ff',p,LCFS} + w_{ip} \mathcal{L}_{ip} + w_{b,MPG} \mathcal{L}_{b,MPG} + w_{b,FLG} \mathcal{L}_{b,FLG} + w_{p,max} (\mathcal{L}_{p,insidemax} + \mathcal{L}_{p,outsidemax})$$

Inside the LCFS: $\mathcal{L}_{p,inside} = \frac{1}{M_p} \sum_{i_p=1}^{M_p} \left[\left(\frac{\partial^2 \psi}{\partial R^2} - \frac{1}{R} \frac{\partial \psi}{\partial R} + \frac{\partial^2 \psi}{\partial Z^2} + \mu_0 R^2 \frac{dp}{d\psi} + \frac{1}{2} \frac{df^2}{d\psi} \right)^2 \right]_{(R_{ip}, Z_{ip})_{inside}}$

$\mathcal{L}_{p,insidemax} \rightarrow$ top max 5% $\mathcal{L}_{p,inside}$

Outside the LCFS: $\mathcal{L}_{p,outside} = \frac{1}{M_p} \sum_{i_p=1}^{M_p} \left[\left(\frac{\partial^2 \psi}{\partial R^2} - \frac{1}{R} \frac{\partial \psi}{\partial R} + \frac{\partial^2 \psi}{\partial Z^2} \right)^2 \right]_{(R_{ip}, Z_{ip})_{outside}}$

$\mathcal{L}_{p,outsidemax} \rightarrow$ top max 5% $\mathcal{L}_{p,outside}$

p' on the LCFS: $\mathcal{L}_{p',p,LCFS} = \frac{1}{M_{p,LCFS}} \sum_{i_{p,LCFS}=1}^{M_{p,LCFS}} \left(p' (R_{i_{p,LCFS}}, Z_{i_{p,LCFS}}) \right)^2$

□ Strong Constraints for points that are difficult to converge

ff' on the LCFS: $\mathcal{L}_{ff',p,LCFS} = \frac{1}{M_{p,LCFS}} \sum_{i_{p,LCFS}=1}^{M_{p,LCFS}} \left(ff' (R_{i_{p,LCFS}}, Z_{i_{p,LCFS}}) \right)^2$

■ Physics loss

Magnetic probes: $\mathcal{L}_{b,MP} = \frac{1}{M_{b,MP}} \sum_{i_{b,MP}=1}^{M_{b,MP}} \frac{\left(B_{Rplasma} (R_{i_{b,MP}}, Z_{i_{b,MP}}) \cos(\theta_{i_{b,MP}}) + B_{Zplasma} (R_{i_{b,MP}}, Z_{i_{b,MP}}) \sin(\theta_{i_{b,MP}}) - B_{plasma}^b_{i_{b,MP}} \right)^2}{(\sigma_{i_{b,MP}}^b)^2}$

$$B_{plasma}^b_{i_{b,MP}} = B_{total}^b_{i_{b,MP}} - G_b J_{ext}$$

$$\mathcal{L}_{b,MPG} = \frac{1}{M_{b,MP}} \sum_{i_{b,MP}=1}^{M_{b,MP}} \frac{\left(G_{MP} I_{pgrid} - B_{plasma}^b_{i_{b,MP}} \right)^2}{(\sigma_{i_{b,MP}}^b)^2}$$

Flux loops: $\mathcal{L}_{b,FL} = \frac{1}{M_{b,FL}} \sum_{i_{b,FL}=1}^{M_{b,FL}} \frac{\left(\psi_{plasma}(R_{i_{b,FL}}, Z_{i_{b,FL}}) - \psi_{plasma}^b_{i_{b,FL}} \right)^2}{(\sigma_{i_{b,FL}}^b)^2}$

$$\mathcal{L}_{b,FLG} = \frac{1}{M_{b,FL}} \sum_{i_{b,FL}=1}^{M_{b,FL}} \frac{\left(G_{FL} I_{pgrid} - \psi_{plasma}^b_{i_{b,FL}} \right)^2}{(\sigma_{i_{b,FL}}^b)^2}$$

$$\psi_{plasma}^b_{i_{b,FL}} = \psi_{total}^b_{i_{b,FL}} - G_f J_{ext}$$

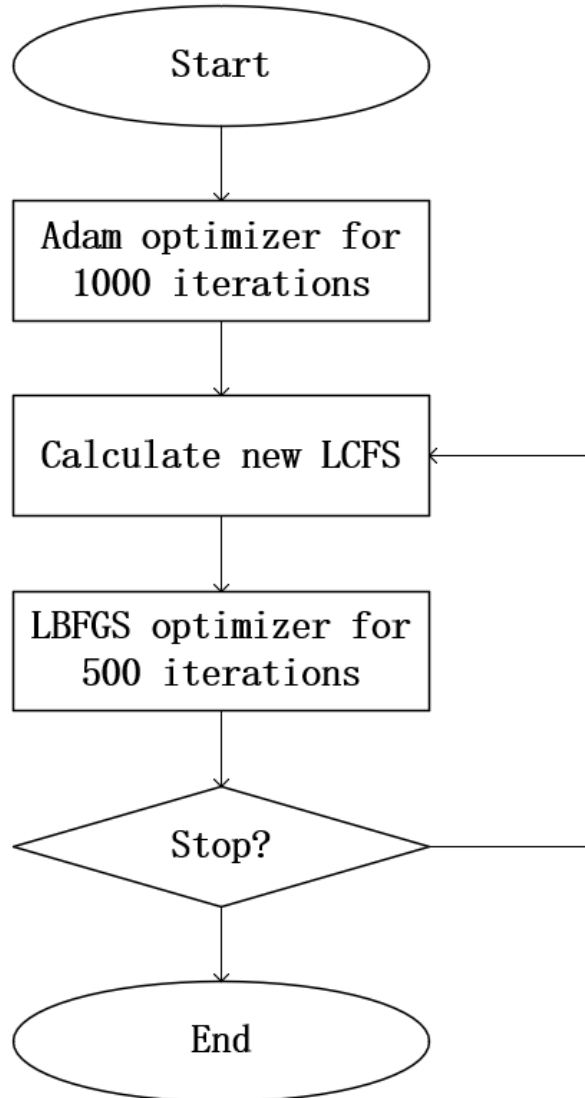
ip: $\mathcal{L}_{ip} = \left(\sum_{i_p=1}^{M_p} I_{pgrid} (R_{ip}, Z_{ip})_{inside} - i_{pmeasure} \right)^2$

$$I_{pgrid} = \left(R p' + \frac{f f'}{\mu_0 R} \right) S_{grid}$$

■ Boundary loss

$$\mathcal{L} = w_{p,inside} \mathcal{L}_{p,inside} + w_{p,outside} \mathcal{L}_{p,outside} + w_{b,MP} \mathcal{L}_{b,MP} + w_{b,FL} \mathcal{L}_{b,FL} + w_{p',p,LCFS} \mathcal{L}_{p',p,LCFS} + w_{ff',p,LCFS} \mathcal{L}_{ff',p,LCFS} + w_{ip} \mathcal{L}_{ip} + w_{b,MPG} \mathcal{L}_{b,MPG} + w_{b,FLG} \mathcal{L}_{b,FLG} + w_{p,max} (\mathcal{L}_{p,insidemax} + \mathcal{L}_{p,outsidemax})$$

Optimizer: Adam, LBFGS

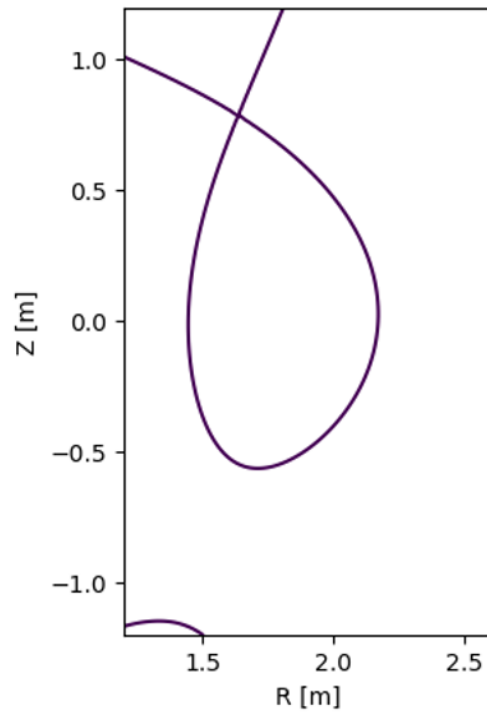


	Adam	LBFGS(below 3500 iterations)	LBFGS(above 3500 iterations)
$w_{p,inside}$	1	10	10
$w_{p,outside}$	1	10	10
$w_{b,MP}$	1	1	1
$w_{b,FL}$	1	1	1
$w_{b,MPG}$	1	10	10
$w_{b,FLG}$	1	10	10
$w_{p',p,LCFS}$	1	10	10
$w_{ff',p,LCFS}$	0.5	5	5
w_{ip}	0.5	1	1
$w_{p,max}$	0	1	10

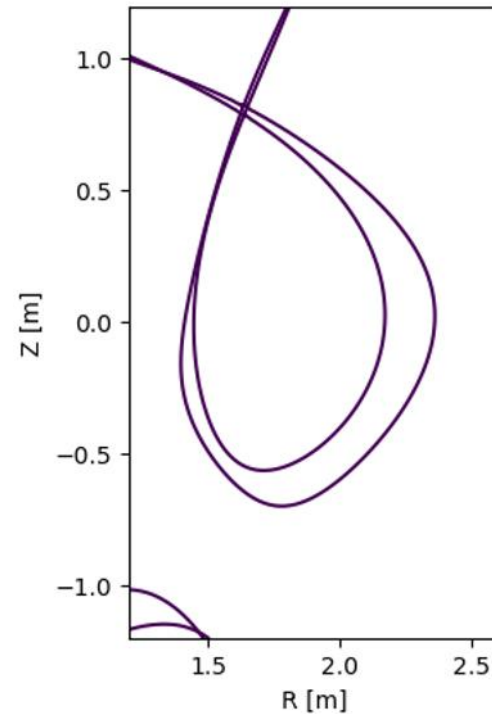
Stop conditions:

- LCFS unchanged between two LBFGS optimization phase
- More than 50,000 iterations

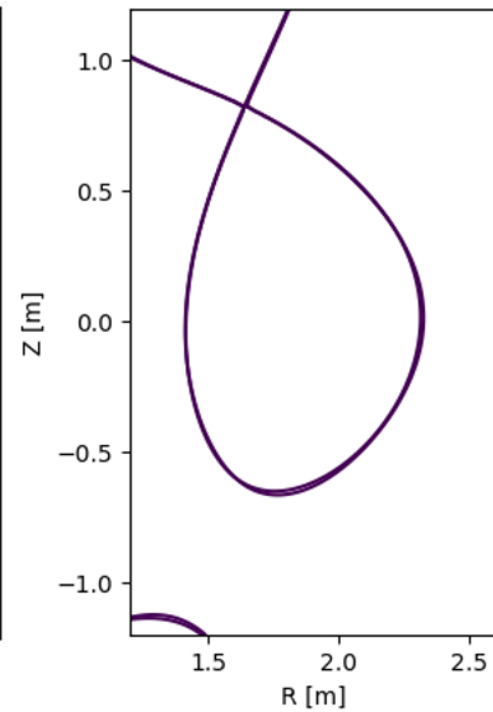
LCFS Convergence Process



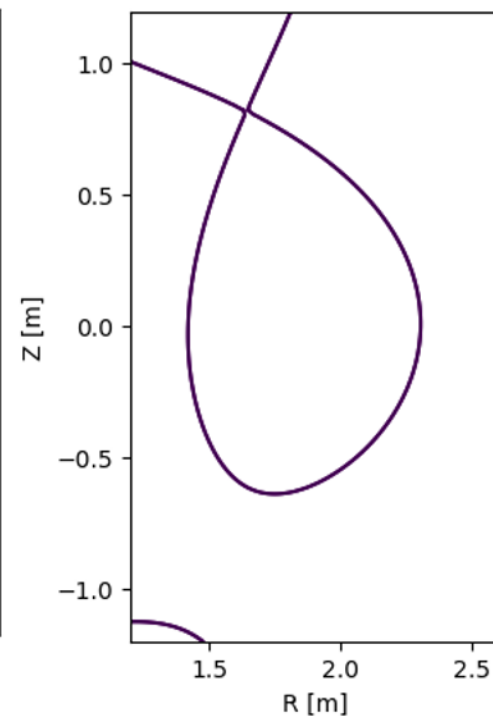
After Adam 1000 iterations



After LBFSGS 500 iterations



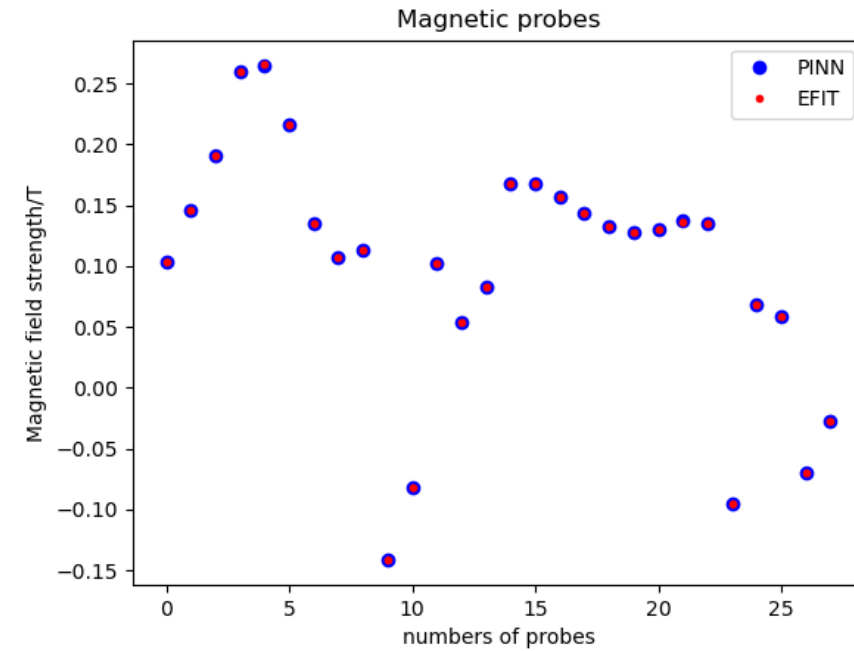
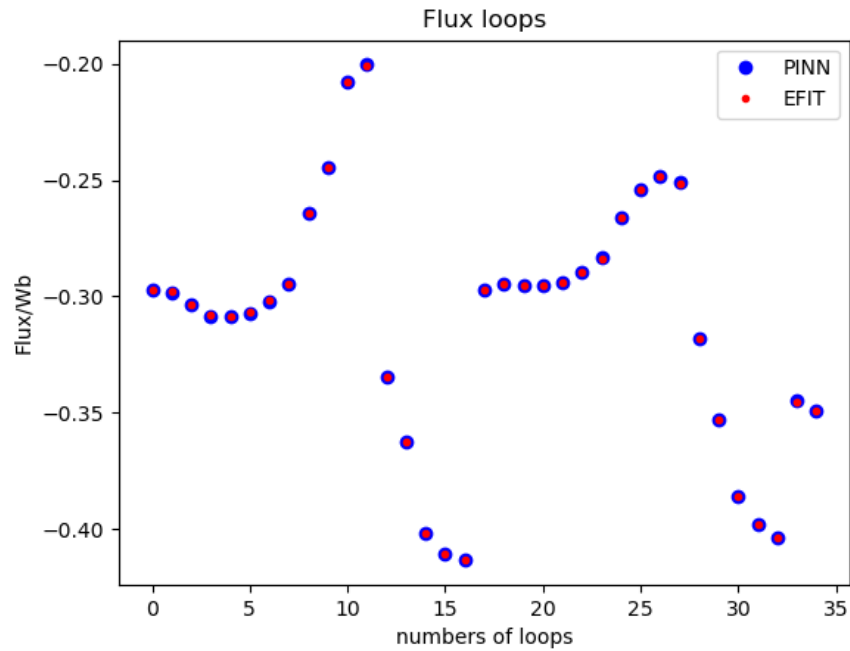
After LBFSGS 2000 iterations



After LBFSGS 4000 iterations

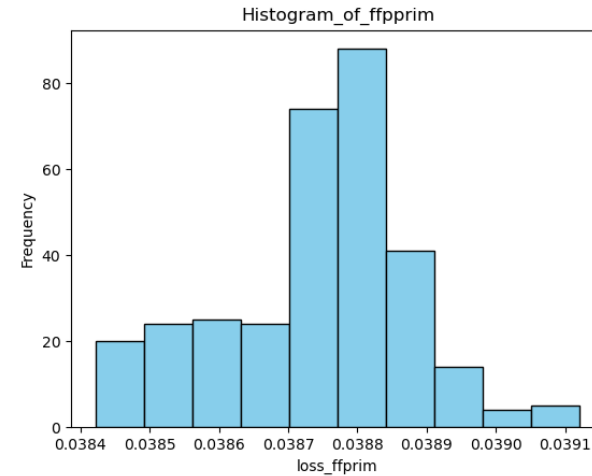
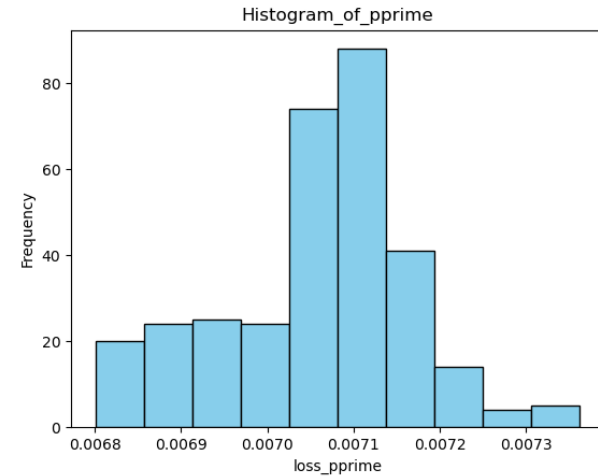
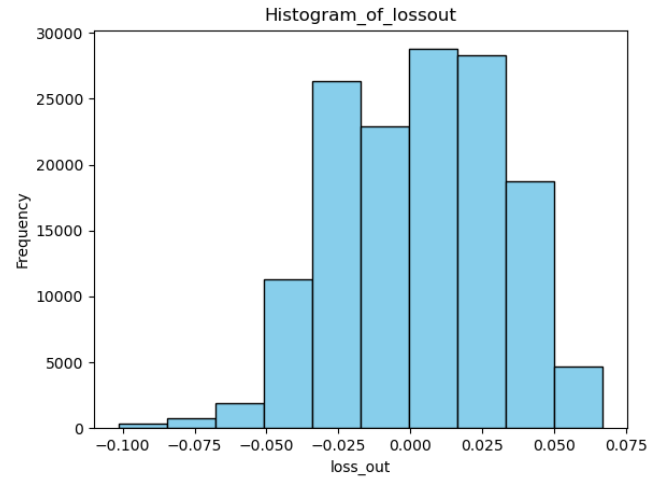
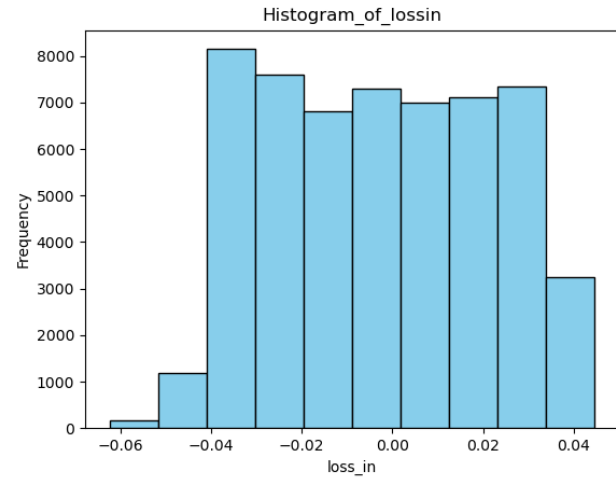
LCFS gradually converges during the optimization process

Convergence of diagnostics



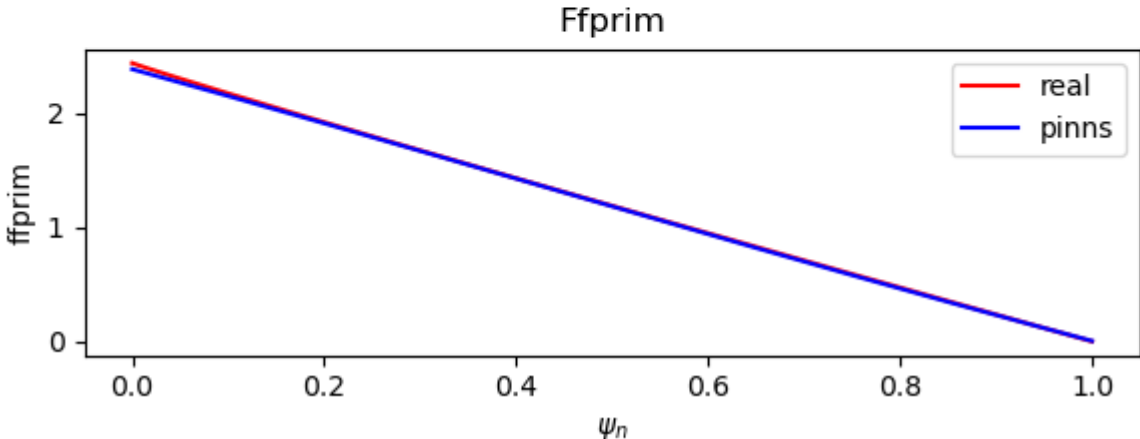
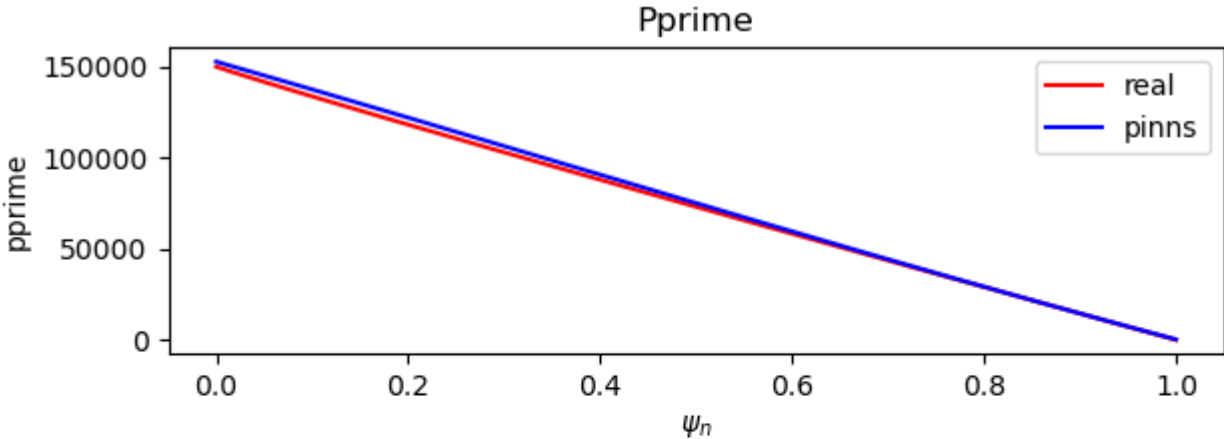
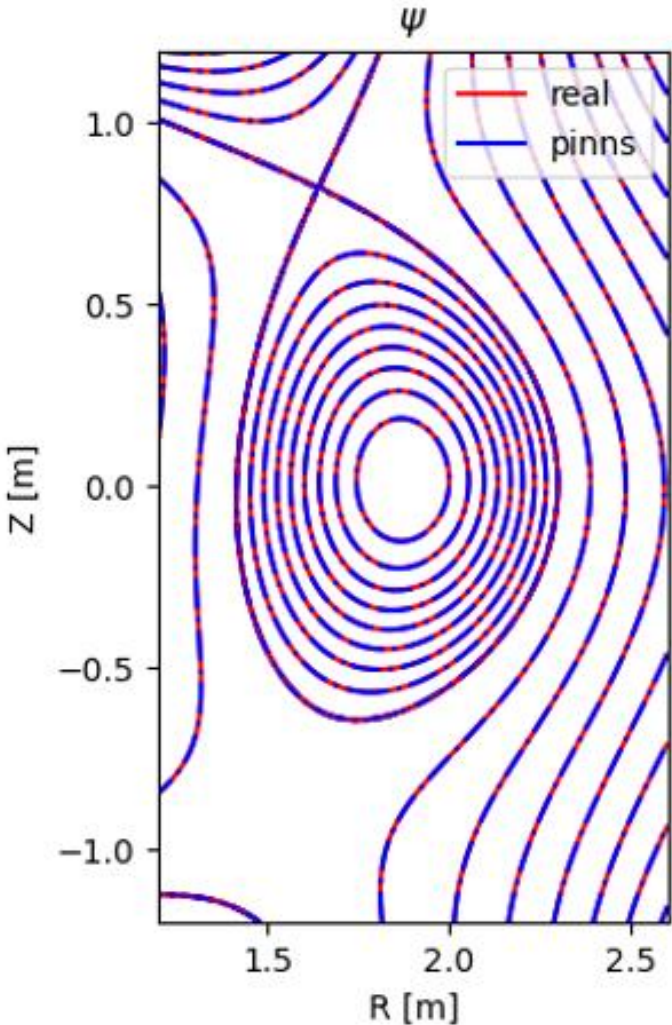
- ✓ Maximum relative error 0.22% on flux loops
- ✓ Maximum relative error 0.26% on magnetic probes

Convergence of physics loss



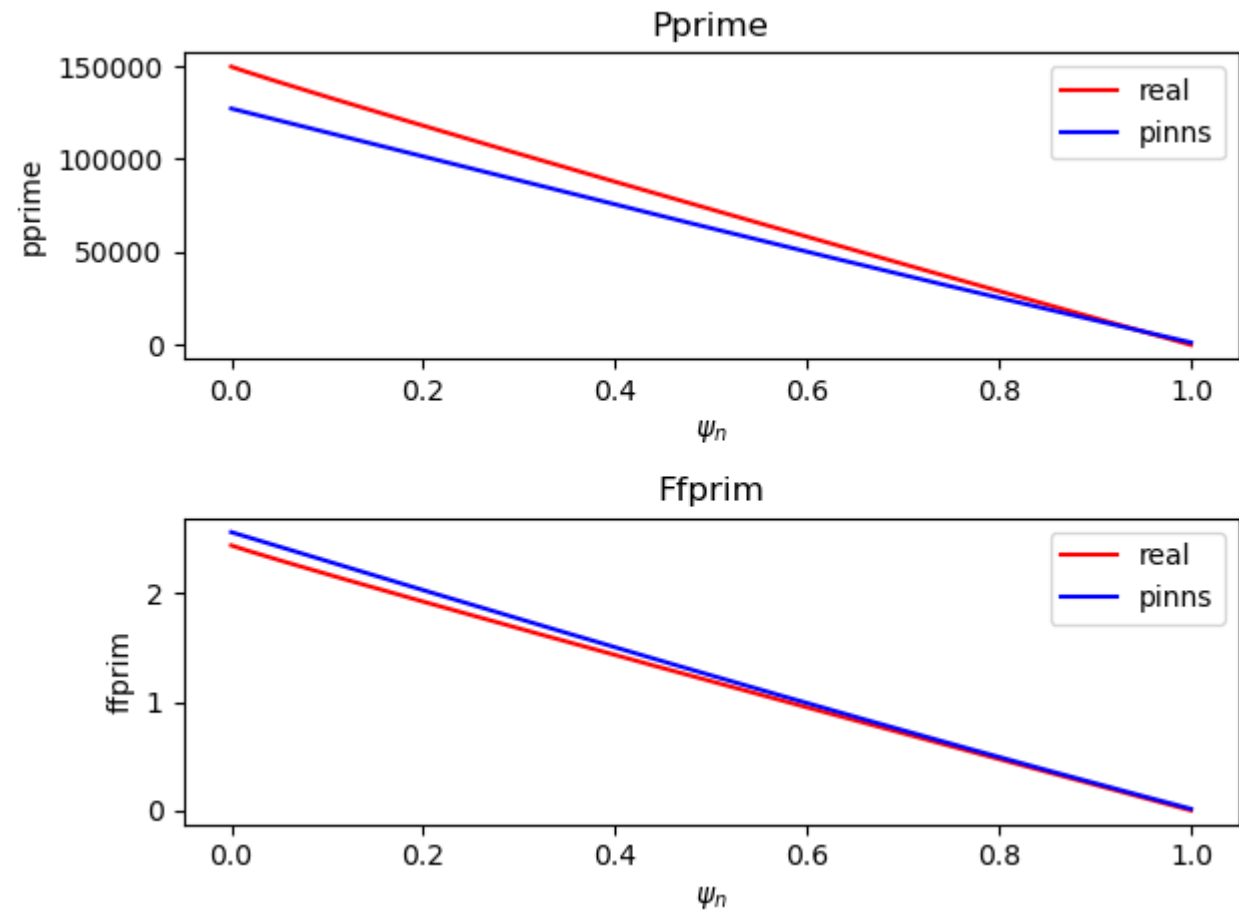
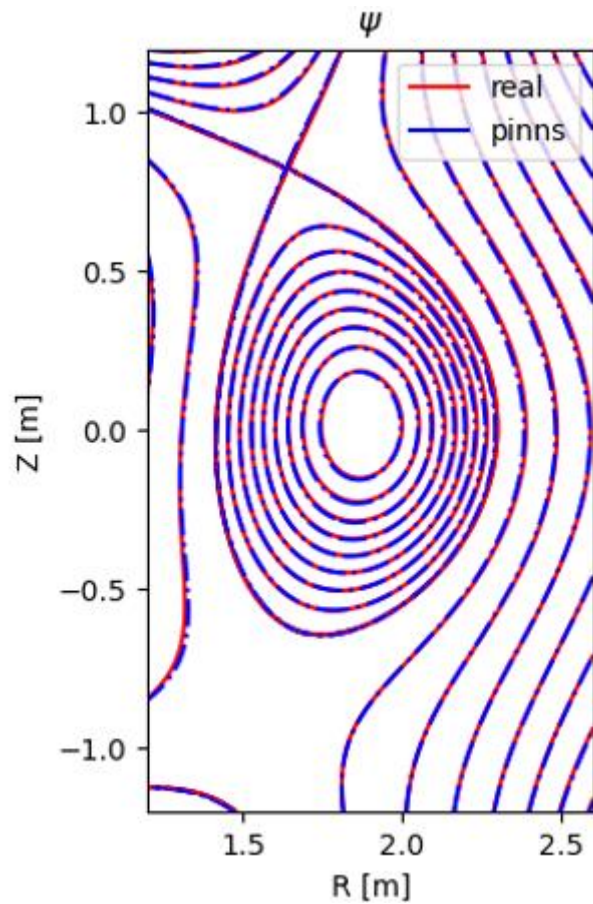
- ✓ an absolute error below 0.06 in LCFS and below 0.1 outside LCFS
- ❑ convergence is still not enough for equilibrium reconstruction

Reconstruction Results



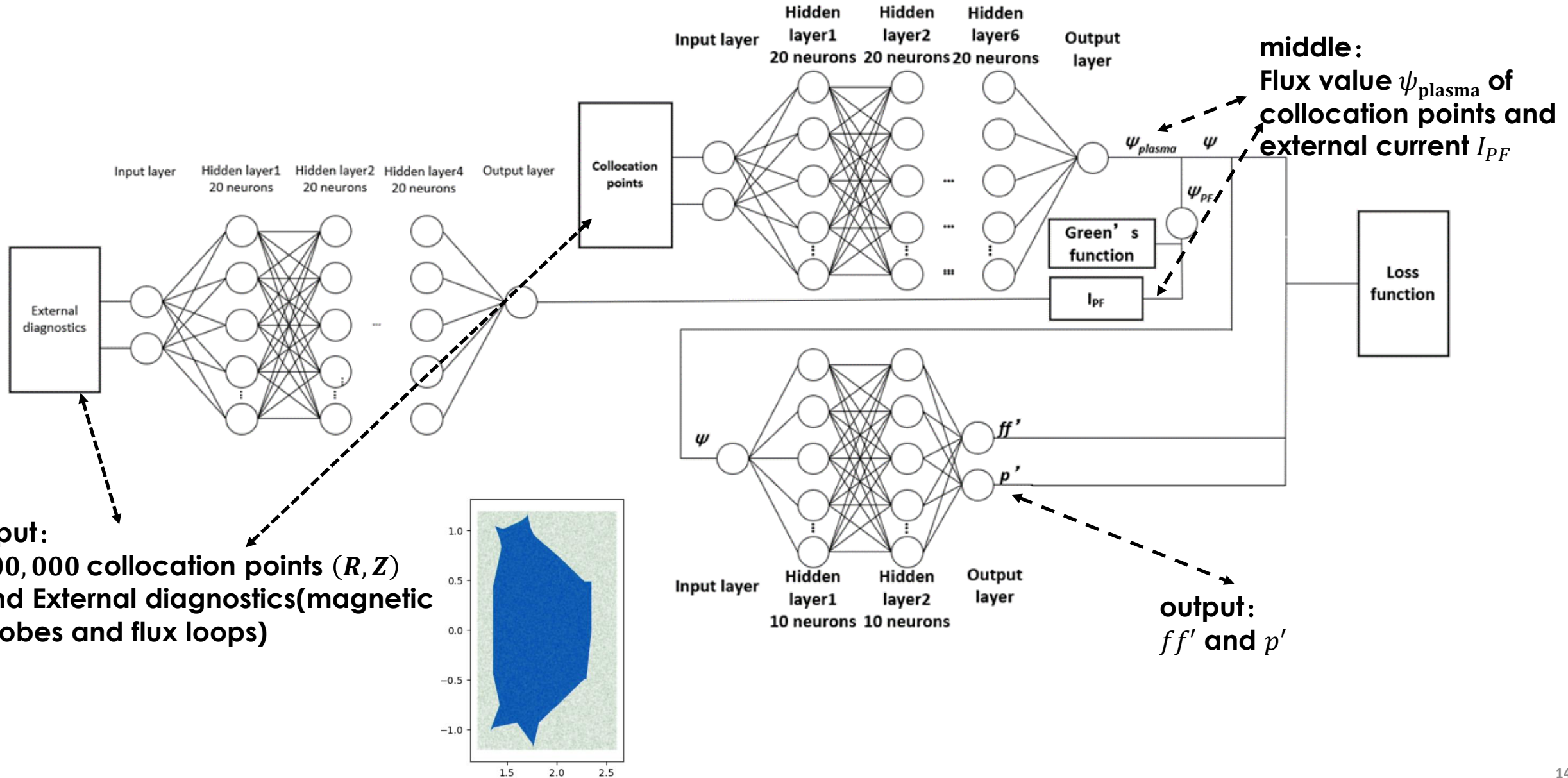
simulated equilibrium reconstruction results
✓ a relative flux error below 0.1%

Reconstruction Results(3% relative diagnostics error)

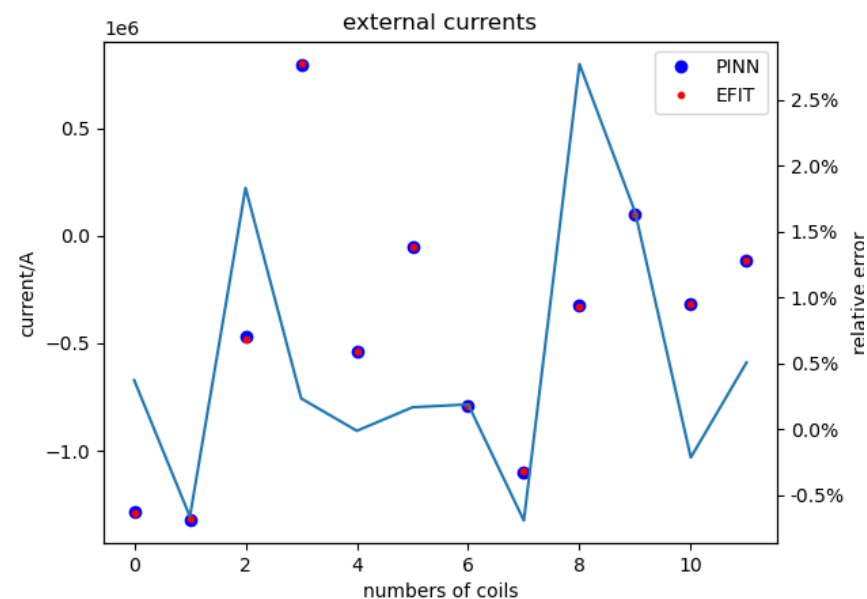
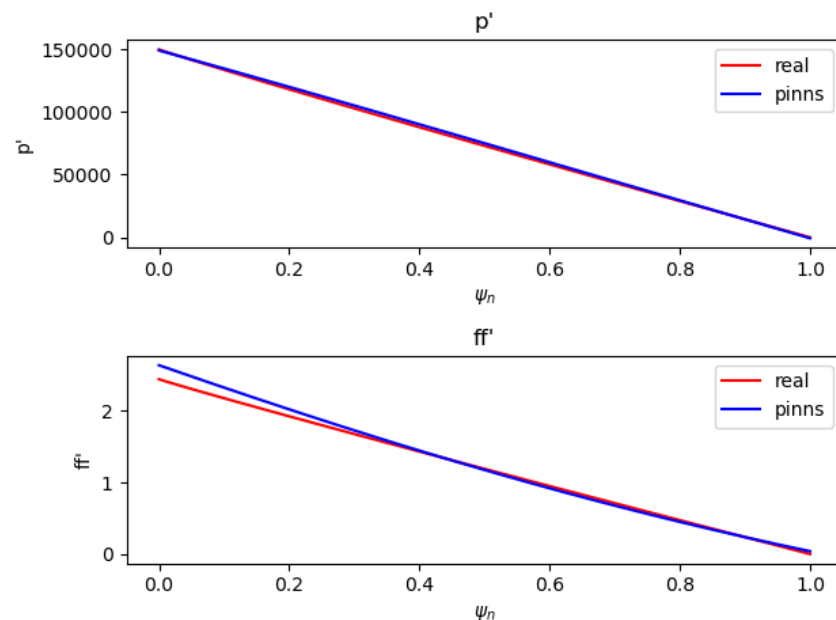
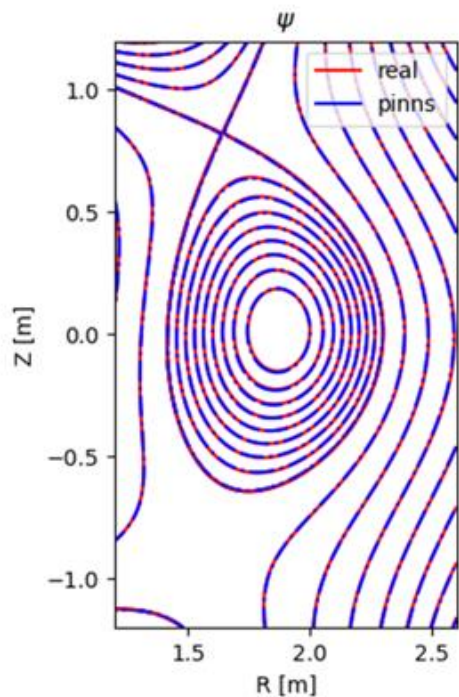


simulated equilibrium reconstruction results (with diagnostics error)

PINN Structure for reconstruction on EAST (I_{PF} unknown)



Reconstruction Results



simulated equilibrium reconstruction results

- ✓ a relative error below 0.5% of flux
- ✓ a relative error below 3% of external currents

□ much harder to converge due to the complexity of the loss function

- ❑ Develop a PINN structure for equilibrium reconstruction with external current known/unknown
- ❑ Design an algorithm, including optimizers and weights adjustment for loss functions
- ✓ The results are all consistent with EFIT, proving its high reliability
- Reconstruction of equilibrium including external currents with relative error added on diagnostics
- Reconstruction of different plasma shapes and complex current distribution form
- Fast equilibrium reconstruction



Thank you !