## TFM PIKAN GPU

June 20, 2025

## [1]: !pip install tensorboard Collecting tensorboard Using cached tensorboard-2.19.0-py3-none-any.whl.metadata (1.8 kB) Collecting absl-py>=0.4 (from tensorboard) Using cached absl\_py-2.3.0-py3-none-any.whl.metadata (2.4 kB) Collecting grpcio>=1.48.2 (from tensorboard) Downloading grpcio-1.73.0-cp312-cp312-win amd64.whl.metadata (4.0 kB) Collecting markdown>=2.6.8 (from tensorboard) Downloading markdown-3.8.2-py3-none-any.whl.metadata (5.1 kB) Requirement already satisfied: numpy>=1.12.0 in c:\users\jorge\anaconda3\envs\kan\_pytorch\lib\site-packages (from tensorboard) (2.0.1)Requirement already satisfied: packaging in c:\users\jorge\anaconda3\envs\kan\_pytorch\lib\site-packages (from tensorboard) (24.2)Collecting protobuf!=4.24.0,>=3.19.6 (from tensorboard) Using cached protobuf-6.31.1-cp310-abi3-win\_amd64.whl.metadata (593 bytes) Requirement already satisfied: setuptools>=41.0.0 in c:\users\jorge\anaconda3\envs\kan\_pytorch\lib\site-packages (from tensorboard) (75.8.0)Requirement already satisfied: six>1.9 in c:\users\jorge\anaconda3\envs\kan\_pytorch\lib\site-packages (from tensorboard) (1.17.0)Collecting tensorboard-data-server<0.8.0,>=0.7.0 (from tensorboard) Using cached tensorboard\_data\_server-0.7.2-py3-none-any.whl.metadata (1.1 kB) Collecting werkzeug>=1.0.1 (from tensorboard) Using cached werkzeug-3.1.3-py3-none-any.whl.metadata (3.7 kB) Requirement already satisfied: MarkupSafe>=2.1.1 in c:\users\jorge\anaconda3\envs\kan\_pytorch\lib\site-packages (from werkzeug>=1.0.1->tensorboard) (2.1.5) Using cached tensorboard-2.19.0-py3-none-any.whl (5.5 MB) Using cached absl\_py-2.3.0-py3-none-any.whl (135 kB) Downloading grpcio-1.73.0-cp312-cp312-win\_amd64.whl (4.3 MB) ----- 0.0/4.3 MB ? eta -:--:------- 4.3/4.3 MB 43.8 MB/s eta 0:00:00 Downloading markdown-3.8.2-py3-none-any.whl (106 kB) Using cached protobuf-6.31.1-cp310-abi3-win amd64.whl (435 kB)

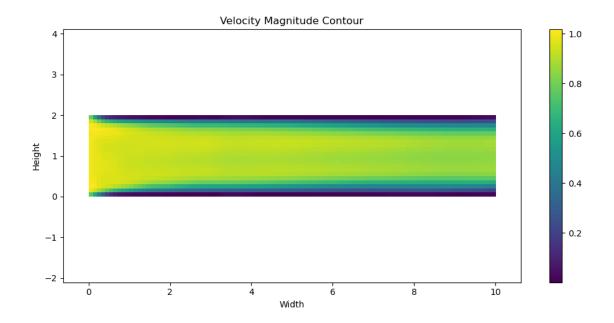
```
Using cached tensorboard_data_server-0.7.2-py3-none-any.whl (2.4 kB)
    Using cached werkzeug-3.1.3-py3-none-any.whl (224 kB)
    Installing collected packages: werkzeug, tensorboard-data-server, protobuf,
    markdown, grpcio, absl-py, tensorboard
    Successfully installed absl-py-2.3.0 grpcio-1.73.0 markdown-3.8.2
    protobuf-6.31.1 tensorboard-2.19.0 tensorboard-data-server-0.7.2 werkzeug-3.1.3
[2]: import torch
     from torch import autograd
     from torch.utils.tensorboard import SummaryWriter
[3]: device = "cuda:0" if torch.cuda.is_available() else "cpu"
[4]: from tqdm import tqdm
     import matplotlib.pyplot as plt
     from kan import KAN, LBFGS
     device = torch.device(device)
     print("Using device:", device)
     rho = torch.tensor(1.0, device=device, requires_grad=False)
     nu = torch.tensor(0.01, device=device, requires_grad=False)
     eps = torch.tensor(1e-8, device=device, requires_grad=False)
     width, height = 10.0, 2.0
     num_points_x, num_points_y = 100, 20
     x = torch.linspace(0, width, num_points_x, device=device, requires_grad=False)
     y = torch.linspace(0, height, num_points_y, device=device, requires_grad=False)
     X, Y = torch.meshgrid(x, y, indexing='ij')
     coordinates = torch.stack([X.flatten(), Y.flatten()], dim=1).to(device)
     coordinates.requires_grad = True # Ensure coordinates require grad
     model = KAN(width=[2,3,3, 3], grid=5, k=10, grid_eps=1.0,
                 noise_scale=0.25).to(device)
     def batch_jacobian(func, x, create_graph=False):
         def _func_sum(x):
            return func(x).sum(dim=0)
         return autograd.functional.jacobian(_func_sum, x,_

¬create_graph=create_graph).permute(1, 0, 2)

     def batch_hessian(func, x):
         jacobian = batch_jacobian(func, x, create_graph=True)
         hessians = []
         for i in range(jacobian.size(1)):
```

```
grad = autograd.grad(jacobian[:, i].sum(), x, create_graph=True,_
 →retain_graph=True) [0]
       hessians.append(grad.unsqueeze(1))
   return torch.cat(hessians, dim=1)
def navier stokes residuals(coords):
    coords = coords.clone().detach().requires_grad_(True) # Ensure coords_u
 →require grad
   y_pred = model(coords)
   grads = batch_jacobian(model, coords, create_graph=True)
   hessians = batch_hessian(model, coords)
   u, v, p = y_pred[:, 0], y_pred[:, 1], y_pred[:, 2]
   u_x, u_y = grads[:, 0, 0], grads[:, 0, 1]
   v_x, v_y = grads[:, 1, 0], grads[:, 1, 1]
   p_x, p_y = grads[:, 2, 0], grads[:, 2, 1]
   u_xx, u_yy = hessians[:, 0, 0], hessians[:, 0, 1]
   v_xx, v_yy = hessians[:, 1, 0], hessians[:, 1, 1]
   continuity = u_x + v_y + eps * p
   x_{momentum} = u * u_x + v * u_y + (1 / rho) * p_x - nu * (u_x + u_y)
   y_{momentum} = u * v_x + v * v_y + (1 / rho) * p_y - nu * (v_x + v_y)
   no_slip_mask = (coords[:, 1] == 0) | (coords[:, 1] == height)
   inlet_mask = (coords[:, 0] == 0)
   outlet mask = (coords[:, 0] == width)
   no_slip_loss = torch.mean(u[no_slip_mask] ** 2 + v[no_slip_mask] ** 2)
   inlet_loss = torch.mean((u[inlet_mask] - 1) ** 2)
   outlet_pressure_loss = torch.mean(p[outlet_mask] ** 2)
   bc_loss = no_slip_loss + inlet_loss + outlet_pressure_loss
   total_loss = torch.mean(continuity ** 2 + x_momentum ** 2 + y_momentum **_u
 42) + bc_loss
   return total_loss
writer = SummaryWriter()
def train():
   optimizer = LBFGS(model.parameters(), lr=1,
                      history_size=10, line_search_fn="strong_wolfe", _
 →tolerance_grad=1e-32, tolerance_change=1e-32, tolerance_ys=1e-32)
    steps = 20 # 20 steps are enough 200 in the original 2h9min version
   pbar = tqdm(range(steps), desc='Training Progress')
```

```
for step in pbar:
             def closure():
                 optimizer.zero_grad()
                 loss = navier_stokes_residuals(coordinates)
                 loss.backward()
                 return loss
             optimizer.step(closure)
             if step \% 5 == 0:
                 current_loss = closure().item()
                 pbar.set_description("Step: %d | Loss: %.3f" %
                                      (step, current_loss))
                 writer.add_scalar('Loss/train', current_loss, step)
     train()
     writer.close()
    Using device: cuda:0
    checkpoint directory created: ./model
    saving model version 0.0
    Step: 15 | Loss: 0.026: 100%|
                                    | 20/20 [05:49<00:00, 17.48s/it]
[5]: u_pred = model(coordinates)[:, 0].detach().reshape(
         num_points_x, num_points_y).T
     v_pred = model(coordinates)[:, 1].detach().reshape(
         num_points_x, num_points_y).T
     magnitude = torch.sqrt(u_pred ** 2 + v_pred ** 2)
     magnitude = magnitude.detach().cpu().numpy()
     plt.figure(figsize=(10, 5)) # Set the figure size as needed
     plt.imshow(magnitude, extent=(0, width, 0, height), origin='lower', __
      ⇔cmap='viridis')
     plt.colorbar() # Add a colorbar to show the magnitude scale
     plt.title('Velocity Magnitude Contour')
     plt.xlabel('Width')
     plt.ylabel('Height')
     plt.axis('equal') # Ensure the plot has equal scaling
     plt.tight_layout() # Adjust layout to prevent overlap
     plt.show()
```

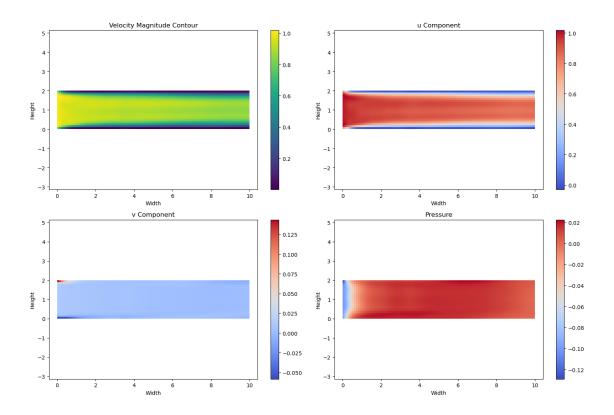


```
[6]: # Extracting predictions
     u_pred = model(coordinates)[:, 0].detach().reshape(num_points_x, num_points_y).T
     v_pred = model(coordinates)[:, 1].detach().reshape(num_points_x, num_points_y).T
     p_pred = model(coordinates)[:, 2].detach().reshape(num_points_x, num_points_y).T
     # Velocity Magnitude
     magnitude = torch.sqrt(u_pred ** 2 + v_pred ** 2)
     magnitude = magnitude.detach().cpu().numpy()
     u_pred = u_pred.detach().cpu().numpy()
     v_pred = v_pred.detach().cpu().numpy()
     p_pred = p_pred.detach().cpu().numpy()
     # Plotting all subplots
     fig, axs = plt.subplots(2, 2, figsize=(15, 10))
     # Velocity Magnitude
     im0 = axs[0, 0].imshow(magnitude, extent=(0, width, 0, height), origin='lower',
     ⇔cmap='viridis')
     fig.colorbar(im0, ax=axs[0, 0])
     axs[0, 0].set_title('Velocity Magnitude Contour')
     axs[0, 0].set_xlabel('Width')
     axs[0, 0].set_ylabel('Height')
     axs[0, 0].axis('equal')
     # u Component
```

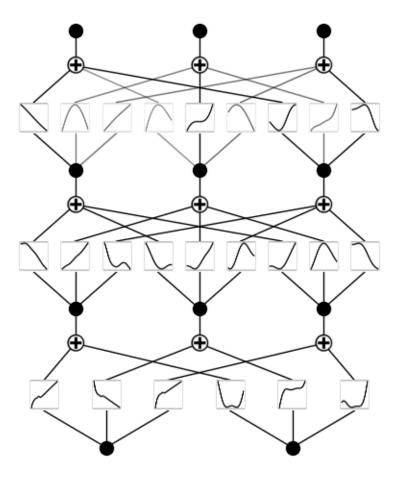
```
im1 = axs[0, 1].imshow(u_pred, extent=(0, width, 0, height), origin='lower',__
 ⇔cmap='coolwarm')
fig.colorbar(im1, ax=axs[0, 1])
axs[0, 1].set_title('u Component')
axs[0, 1].set_xlabel('Width')
axs[0, 1].set ylabel('Height')
axs[0, 1].axis('equal')
# v Component
im2 = axs[1, 0].imshow(v_pred, extent=(0, width, 0, height), origin='lower', __
⇔cmap='coolwarm')
fig.colorbar(im2, ax=axs[1, 0])
axs[1, 0].set_title('v Component')
axs[1, 0].set_xlabel('Width')
axs[1, 0].set_ylabel('Height')
axs[1, 0].axis('equal')
# Pressure
im3 = axs[1, 1].imshow(p_pred, extent=(0, width, 0, height), origin='lower', __

cmap='coolwarm')

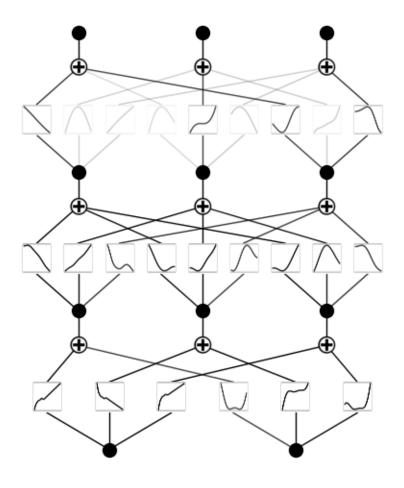
fig.colorbar(im3, ax=axs[1, 1])
axs[1, 1].set_title('Pressure')
axs[1, 1].set_xlabel('Width')
axs[1, 1].set_ylabel('Height')
axs[1, 1].axis('equal')
plt.tight_layout() # Adjust layout to prevent overlap
plt.show()
```



## [7]: model.plot(beta=10)



[8]: model.plot()



## [9]: model.auto\_symbolic()

```
fixing (0,0,0) with x, r2=0.9735497236251831, c=1
fixing (0,0,1) with 0, r2=0.0, c=0
fixing (0,0,2) with 0, r2=0.0, c=0
fixing (0,1,0) with 0, r2=0.0, c=0
fixing (0,1,1) with 0, r2=0.0, c=0
fixing (0,1,2) with 0, r2=0.0, c=0
fixing (1,0,0) with sin, r2=0.9986881613731384, c=2
fixing (1,0,1) with x, r2=0.992422342300415, c=1
fixing (1,0,2) with 0, r2=0.0, c=0
fixing (1,1,0) with 0, r2=0.0, c=0
fixing (1,1,1) with 0, r2=0.0, c=0
fixing (1,1,2) with cos, r2=0.9962447285652161, c=2
fixing (1,2,0) with 0, r2=0.0, c=0
fixing (1,2,1) with 0, r2=0.0, c=0
fixing (1,2,2) with \sin, r2=0.9975288510322571, c=2
fixing (2,0,0) with \sin, r2=1.000000238418579, c=2
fixing (2,0,1) with \sin, r2=0.9994560480117798, c=2
```

```
fixing (2,0,2) with x^2, r2=0.999975860118866, c=2
      fixing (2,1,0) with 0, r2=0.0, c=0
      fixing (2,1,1) with 0, r2=0.0, c=0
      fixing (2,1,2) with cos, r2=0.9997799396514893, c=2
      fixing (2,2,0) with 0, r2=0.0, c=0
      fixing (2,2,1) with 0, r2=0.0, c=0
      fixing (2,2,2) with 0, r2=0.0, c=0
      saving model version 0.1
[10]: from kan.utils import ex_round
[11]: formula1, formula2, formula3 = model.symbolic_formula()[0]
[12]: ex_round(formula1,2)
[12]: 1.08\sin{(1.71\sin{(0.17x_1-4.46)}+8.52)}+0.48
[13]: ex_round(formula2,2)
[13] : -0.01\sin{(2.96\sin{(0.17x_1-4.46)}-9.56)}
[14]: ex_round(formula3,2)
 \begin{array}{l} \textbf{[14]:} \\ -0.03 \left(1-0.42 \sin \left(0.17 x_1-4.46\right)\right)^2-0.05 \cos \left(0.12 x_1+8.95\right)-0.01 \end{array}
```