summarize

May 16, 2023

1

matplotlib

1.1 Loss

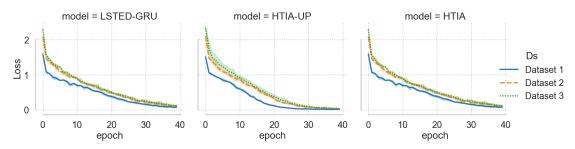
```
[2]: # loss GRu
    import os
    import pandas as pd
    import matplotlib.pyplot as plt
    # excel
    dir_path1= r'lstm_attn\
                              \sad'
    dir_path2= r' \lstm_attn\ \sad'
    dir_path3= r' \ gru\ \sad' # USED-up Loss3
    def input_data(dir_path):
        file_names = os.listdir(dir_path)
              loss
        all_losses = {}
        count = 0
        group_dict = {}
              excel
                      loss
                               epoch
        for file_name in file_names:
            file_path = os.path.join(dir_path, file_name)
            df = pd.read_excel(file_path)
            #
            if count % 5 == 0:
                group_name = 'Dataset {}'.format(int(count/5)+1)
                group_dict[group_name] = []
               epoch loss
```

```
model_name = file_name.split(".")[0]
                   epoch_losses = []
                   for epoch in range(0, 100):
                             batch_losses = df[df['Step'] == epoch]['Value'].tolist()
                             epoch_loss = sum(batch_losses) / len(batch_losses)
                             epoch_losses.append(epoch_loss)
                   # loss
                   group dict[group name].append(epoch losses)
                   count += 1
                                         loss
         for group_name, group_losses in group_dict.items():
                   epoch_losses_avg = []
                   for epoch in range(0, 100):
                             epoch_losses_sum = 0
                             for model_losses in group_losses:
                                      epoch_losses_sum += model_losses[epoch]
                             epoch_losses_avg.append(epoch_losses_sum/len(group_losses))
                   all_losses[group_name] = epoch_losses_avg
              fig, ax = plt.subplots(figsize=(8, 6), dpi=300)
         colors = ['r', 'g', 'b']
         data list = []
         label_list = []
         for i, (group_name, group_losses) in enumerate(group_dict.items()):
                   for j, model_losses in enumerate(group_losses):
                                  ax.plot(model\_losses[0:40], color=colors[i], label='{}({})'.
   \hookrightarrow format(group_name, j+1))
                             data_list.append(model_losses[0:40])
                            label_list.append('{}'.format(group_name))
                                  label_list.append('{}({})'.format(group_name, j+1))
         df = pd.DataFrame(data_list).T
         df.columns = label list
         df['epoch']=df.index
         df=df.melt(id_vars=['epoch'],
                   value_vars=['Dataset 1', 'Dataset 1', '
                 'Dataset 2', 'Dataset 2', 'Dataset 2', 'Dataset 2', 'Dataset 2',
                 'Dataset 3', 'Dataset 3', 'Dataset 3', 'Dataset 3'],
                  var_name=["Ds"],
                  value_name="Loss"
                 )
         return df
df1=input_data(dir_path1)
df2=input_data(dir_path2)
df3=input_data(dir_path3)
```

```
data=pd.concat([df2,df3,df1],keys=['LSTED-GRU','HTIA-UP','HTIA'])
data['model']=data.index.get_level_values(0)
```

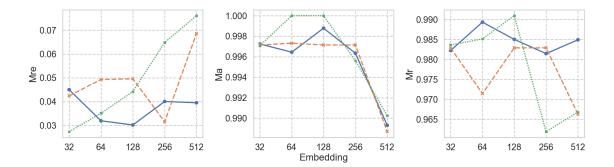
```
[3]: import seaborn as sns
     sns.set_style("whitegrid", {
         "grid.linestyle": ":",
         "lines.solid_capstyle":'round',
     })
     sns.set_context('poster')
     # Plot the responses for different events and regions
     plt.figure(figsize=(15, 10))
     g=sns.relplot(x="epoch", y="Loss",
                  hue="Ds", style="Ds", kind='line',
                   errorbar="sd",
                 col='model',
                  data=data)
     g.add_legend(frameon=True)
     g.legend.set_bbox_to_anchor((1.1, 0.5))
     sns.despine(offset=2, trim=True)
     plt.savefig("pdf\Loss_plot.pdf",dpi=600)
```

<Figure size 1500x1000 with 0 Axes>



1.2

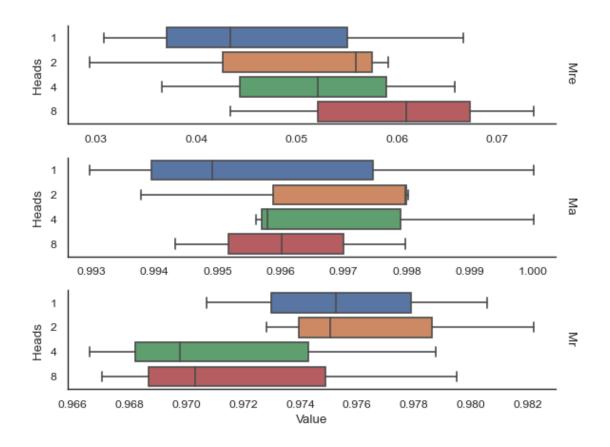
```
sns.set_style("whitegrid")
plt.figure(figsize=(20,6))
sns.set_context('poster')
pc = sns.color_palette('deep', 3)
plt.rcParams["grid.linestyle"] = "--"
plt.rcParams['axes.linewidth'] = 2
plt.subplot(1,3,1)
# sns.set_context('poster')
sns.lineplot(x='embedding', y='Mre',palette=pc ,
             hue='data',data=data, legend=False, style='data')
sns.scatterplot(x='embedding', y='Mre',palette=pc, s=100,
                hue='data', data=data, legend=False, style='data')
plt.xlabel("")
plt.subplot(1,3,2)
sns.lineplot(x='embedding', y='Ma',palette=pc ,
             hue='data',data=data, legend=False, style='data')
sns.scatterplot(x='embedding', y='Ma',palette=pc, s=100,
                hue='data', data=data, legend=False, style='data')
plt.xlabel("Embedding")
plt.subplot(1,3,3)
sns.lineplot(x='embedding', y='Mr',palette=pc ,
             hue='data',data=data, legend=False, style='data')
sns.scatterplot(x='embedding', y='Mr',palette=pc, s=100,
                hue='data', data=data, legend=False, style='data')
# plt.legend(loc='center left', bbox_to_anchor=(1.05, 0.5))
plt.xlabel("")
plt.tight_layout() #
plt.savefig ( "pdf/embedding.pdf", dpi= 600)
```



1.3

\mathbf{Z}

```
[34]: import seaborn as sns
      import pandas as pd
      import matplotlib.pyplot as plt
      data = pd.read csv("
                           \head.csv")
      data['head']=['1','2','4','8']*3
      data=data[['data','head', 'Mre3', 'Ma3', 'Mr3',]]
      df=data.melt(id_vars=['data', 'head'],
              value_vars=['Mre3', 'Ma3', 'Mr3',],
              var_name="Index",
              value_name="Value"
      df['Index'] = df['Index'].str.split("3",expand=True)[0]
      # plt.figure(figsize=(20,12))
      sns.set_context('paper')
      g = sns.catplot(
          data=df,
          x="Value", y="head", row="Index",legend='strip',legend_out=True,
          kind="box", orient="h",
          sharex=False, margin_titles=True,
          height=1.5, aspect=4,
      g.set(ylabel="Heads ")
      g.set_titles(row_template="{row_name} ")
      plt.savefig('pdf\plotheading.pdf', dpi=300)
```



1.4 chord

d3block

```
import seaborn as sns
import matplotlib.pyplot as plt

import pickle
import matplotlib.colors as mcolors
import os
os.environ["KMP_DUPLICATE_LIB_OK"]="TRUE"

import numpy as np
import pandas as pd
import torch
with open(" \\attn\\ numpairs attr\\attn_2.pickle", "rb") as file:
    attn2= pickle.load( file).cpu()
with open(" \\attn\\ numpairs attr\\env2.pickle", "rb") as file:
    environment2= pickle.load( file)
data2=attn2[:,0,:,:].squeeze(0).numpy() #
```

```
numpairs,attr,label=environment2
numpair_numpy=numpairs.squeeze(0).cpu().numpy().astype(str)
dataframe2=pd.DataFrame(data2,index=numpair_numpy,columns=numpair_numpy)
new_df = dataframe2.drop(index='0',columns='0').stack().reset_index()
new_df.columns = ['source', 'target', 'weight']
new_df
```

```
[145]:
          source target
                             weight
                     9 0.00000e+00
      0
               9
               9
                     5 5.000000e-01
      1
      2
               9
                     5 5.000000e-01
      3
               9
                     6 2.207841e-35
                     6 2.207841e-35
      4
               9
                     4 0.000000e+00
      1439
               4
      1440
              4
                    4 0.000000e+00
              4
      1441
                     6 0.000000e+00
      1442
             4
                    6 0.000000e+00
      1443
              4
                     4 0.000000e+00
```

[1444 rows x 3 columns]

$2 \quad ATTN$

attn

```
[32]: import numpy as np
      import pandas as pd
      import seaborn as sns
      import matplotlib.pyplot as plt
      import pickle
      import torch
      def open_attn(data,manu_seed,index):
          with open(rf"lstm_attn\attn\attn {index}_{data}_{manu_seed}.pickle", "rb")__
       ⇔as file:
             attn= pickle.load( file).cpu()
          with open(rf"lstm_attn\attn\env_{index}_{data}_{manu_seed}.pickle", "rb")__
       →as file:
             environment= pickle.load( file)
          with open(rf"lstm_attn\attn\pred_{index}_{data}_{manu_seed}.pickle", "rb")__
       →as file:
              pred= pickle.load(file)
          data=attn[:,0,:,:].squeeze(0).numpy() #
          numpairs, attr, label=environment
          numpair_numpy=numpairs.squeeze(0).numpy()#.astype(str)
```

```
dataframe=pd.DataFrame(data,index=numpair_numpy,columns=numpair_numpy)
   return attn, numpairs, attr, label, pred
def group_tensors_by_value(tensor_list):
       tensor
                        tensor
    :param tensor_list: tensor
    :return: tensor
    11 11 11
          tensor
                          tensor
   tensor dict = {}
   for i, tensor in enumerate(tensor_list):
        found = False
        for key in tensor_dict:
            if torch.equal(key, tensor):
                tensor_dict[key].append(i)
                found = True
                break
        if not found:
            tensor_dict[tensor] = [i]
                  tensor
   grouped_tensors = []
   index list = []
   for tensor, indices in tensor_dict.items():
        if len(indices) > 1:
            grouped_tensor = torch.stack([tensor_list[i] for i in indices])
            grouped_tensors.append(grouped_tensor)
            index_list.append(indices)
   return grouped_tensors, index_list
def attn_clean(attn,numpairs):
   numpair_numpy=numpairs.squeeze(0).cpu().numpy().astype(str)
   dataframe=pd.DataFrame(attn.squeeze(0)[0,:,:
 -, index=numpair_numpy, columns=numpair_numpy).drop(index='0', columns='0')
   return dataframe
def get_data(Ds,select_datas,length):
   manu_seeds = [f"{i}" for i in range(5)]
   txtpaths = ["981762","981808","981814"]
   numpair_list=[open_attn(Ds,i,j)[1] for i in manu_seeds for j in_
 →range(length)]
   raw_list=[open_attn(Ds,i,j) for i in manu_seeds for j in range(length)]
```

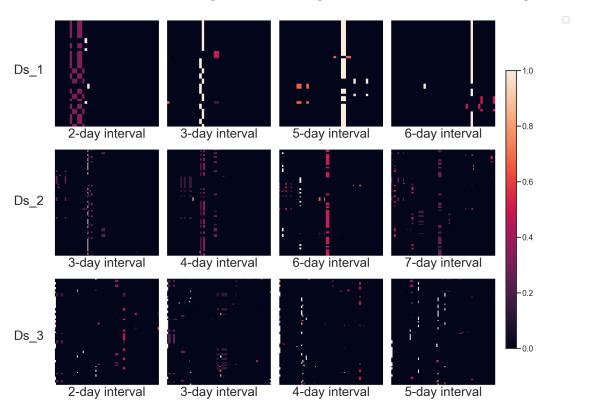
```
_,index=group_tensors_by_value(numpair_list)
    pred=[raw_list[i][-1] for i in select_datas]
    attr=[raw_list[i][-3] for i in select_datas]
    print(pred,attr)
    datas=[attn_clean(*raw_list[i][0:2]) for i in select_datas]
    return datas
ds1=get_data('981762',[ 163,136,106,68],164)
ds2=get_data('981808',[20, 65, 242, 246],70)
ds3=get_data('981814',[ 550, 492,152, 182 ],114)
datas=ds1+ds2+ds3
[[0.7, 1.0, 0.6], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]]
[tensor([[[3, 8],
         [4, 7],
         [5, 6],
         [6, 5],
         [0, 4],
         [1, 3],
         [2, 2]]]), tensor([[[3, 9],
         [4, 8],
         [5, 7],
         [6, 6],
         [0, 5],
         [1, 4],
         [2, 3]]]), tensor([[[ 3, 11],
         [4, 10],
         [5, 9],
         [6, 8],
         [0, 7],
         [1, 6],
         [ 2, 5]]]), tensor([[[ 3, 12],
         [4, 11],
         [5, 10],
         [6, 9],
         [0, 8],
         [1, 7],
         [2, 6]]])]
[[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.09090909090909091, 1.0,
0.90909090909091], [0.0, 1.0, 1.0]] [tensor([[[3, 9],
         [4, 8],
         [5, 7],
         [6, 6],
         [0, 5],
         [1, 4],
         [2, 3]]]), tensor([[[ 3, 10],
         [4, 9],
         [5, 8],
```

```
[6, 7],
              [0, 6],
              [1, 5],
              [ 2, 4]]]), tensor([[[ 3, 11],
              [4, 10],
              [5, 9],
              [6, 8],
              [0, 7],
              [1, 6],
              [ 2, 5]]]), tensor([[[ 3, 12],
              [4, 11],
              [5, 10],
              [6, 9],
              [0, 8],
              [1, 7],
              [2, 6]]])]
     [[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]]
     [tensor([[[6, 7],
              [0, 6],
              [1, 5],
              [2, 4],
              [3, 3],
              [4, 2],
              [5, 1]]]), tensor([[[6, 8],
              [0, 7],
              [1, 6],
              [2, 5],
              [3, 4],
              [4, 3],
              [5, 2]]]), tensor([[[6, 9],
              [0, 8],
              [1, 7],
              [2, 6],
              [3, 5],
              [4, 4],
              [5, 3]]]), tensor([[[ 6, 11],
              [0, 10],
              [1, 9],
              [2, 8],
              [3, 7],
              [4, 6],
              [5, 5]])]
[36]: import numpy as np
      import seaborn as sns
      import matplotlib.pyplot as plt
```

```
data = np.random.rand(10, 10)
  2x5
             5x5
fig, axes = plt.subplots(3, 4, figsize=(20, 15),)
axes = axes.flatten()
for i, ax in enumerate(axes):
   sns.set_theme(style="ticks")
    sns.set_context("talk")
   sns.set_context("poster")
   sns.color_palette("rocket",as_cmap=True)# as_cmap=True
    sns.heatmap(datas[i], ax=ax, cbar=False, xticklabels=False,
 ⇔yticklabels=False)
     ax.set title(f"Plot ") #
day=[2,3,5,6,3,4,6,7,2,3,4,5]
for i,ax in enumerate(axes):
   ax.set_xticklabels([])
   ax.set_yticklabels([])
   ax.set_xlabel(f"{day[i]}-day interval",fontsize=35)
     ax.set_ylabel("b")
fig.legend()
fig.subplots_adjust(top=0.9, wspace=0.3, hspace=0)
# fig.text(0.5, 1, "X-Axis Title", ha="center", fontsize=22)
# fig.text(-0.02, 0.5, "Y-Axis Title", va="center", rotation="vertical",
 \rightarrow fontsize=22)
fig.text(-0.06, 0.52, "Ds 2", va="center", fontsize=35)
fig.text(-0.06, 0.18, "Ds_3", va="center", fontsize=35)
fig.text(-0.06, 0.85, "Ds_1", va="center", fontsize=35)
# fig.text(0.15, 0.99, "Incidence attention weights when predicting future"
 ⇔results at different time intervals", va="center", fontsize=22)
plt.tight_layout()
fig.subplots_adjust(right=0.85) #
cbar ax = fig.add axes([0.87, 0.15, 0.02, 0.7]) #
fig.colorbar(axes[0].collections[0], cax=cbar_ax) #
plt.savefig('pdf\\attn1.pdf', dpi=900)
```

```
plt.show()
```

No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.



```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import pickle
import torch

def open_attn(data,manu_seed,index):
    with open(rf"lstm_attn\attn\attn_{index}_{data}_{manu_seed}.pickle","rb")
    as file:
        attn= pickle.load( file).cpu()
    with open(rf"lstm_attn\attn\env_{index}_{data}_{manu_seed}.pickle", "rb")
    as file:
        environment= pickle.load( file)
    with open(rf"lstm_attn\attn\env_{index}_{data}_{manu_seed}.pickle", "rb")
    as file:
        environment= pickle.load( file)
    with open(rf"lstm_attn\attn\pred_{index}_{data}_{manu_seed}.pickle", "rb")
    as file:
```

```
pred= pickle.load(file)
   data=attn[:,0,:,:].squeeze(0).numpy() #
   numpairs, attr, label=environment
   numpair_numpy=numpairs.squeeze(0).numpy()#.astype(str)
   dataframe=pd.DataFrame(data,index=numpair_numpy,columns=numpair_numpy)
   return attn, numpairs, attr, label, pred
def group_tensors_by_value(tensor_list):
    11 11 11
                        tensor
       tensor
    :param\ tensor\_list: tensor
    :return: tensor
    11 11 11
           tensor
                         tensor
   tensor_dict = {}
   for i, tensor in enumerate(tensor_list):
        found = False
        for key in tensor_dict:
            if torch.equal(key, tensor):
                tensor_dict[key].append(i)
                found = True
                break
        if not found:
            tensor dict[tensor] = [i]
                   tensor
   grouped_tensors = []
   index_list = []
   for tensor, indices in tensor_dict.items():
        if len(indices) > 1:
            grouped_tensor = torch.stack([tensor_list[i] for i in indices])
            grouped_tensors.append(grouped_tensor)
            index_list.append(indices)
   return grouped_tensors, index_list
def attn_clean(attn,numpairs):
   numpair_numpy=numpairs.squeeze(0).cpu().numpy().astype(str)
   dataframe=pd.DataFrame(attn.squeeze(0)[0,:,:
 →],index=numpair_numpy,columns=numpair_numpy).drop(index='0',columns='0')
   return dataframe
def get_data(Ds,select_datas,length):
   manu seeds = [f"{i}" for i in range(5)]
```

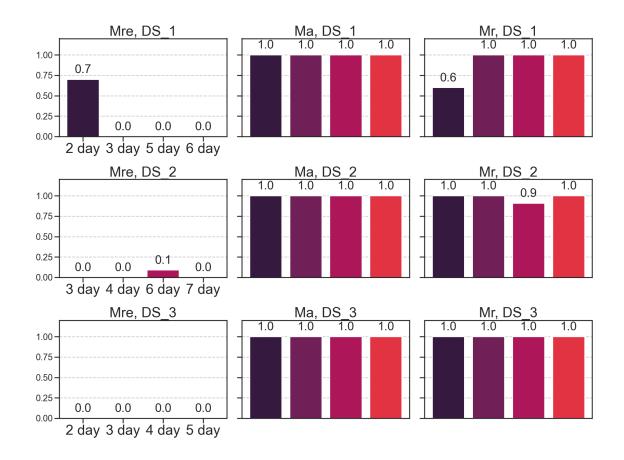
```
txtpaths = ["981762","981808","981814"]
           numpair_list=[open_attn(Ds,i,j)[1] for i in manu_seeds for j in_
        →range(length)]
           raw list=[open attn(Ds,i,j) for i in manu seeds for j in range(length)]
           _,index=group_tensors_by_value(numpair_list)
           pred=[raw list[i][-1] for i in select datas]
           attr=[raw_list[i][-3] for i in select_datas]
           print(pred,attr)
           datas=[attn_clean(*raw_list[i][0:2]) for i in select_datas]
           return datas
       \# manu_seeds = [f''manu_seed=\{i\}'' for i in range(5)]
       # txtpaths = ["981762", "981808", "981814"]
       \# \ q = [[a:1]]
       # [q[-1].append(y)] if x == y else q.append([y]) for x, y in zip(a[:-1], a[1:])
[147]: # 762 [ 163,136,106,68] [[0.7, 1.0, 0.6], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0],
       →[0.0, 1.0, 1.0]] 2356
       # 808 [20, 65, 242, 246] [[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.
       409090909090909091, 1.0, 0.90909090909091],[0.0, 1.0, 1.0]] 3467
       # 814 [ 550, 492,152, 182 ] [[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.
       ⊶0], [0.0, 1.0, 1.0]], 2345
       # 762-->164 808-->70 814-->114
       ds1=get_data('981762',[ 163,136,106,68],164)
       ds2=get_data('981808',[20, 65, 242, 246],70)
       ds3=get_data('981814',[ 550, 492,152, 182 ],114)
       datas=ds1+ds2+ds3
      [[0.7, 1.0, 0.6], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]]
      [tensor([[[3, 8],
               [4, 7],
               [5, 6],
               [6, 5],
               [0, 4],
               [1, 3],
               [2, 2]]]), tensor([[[3, 9],
               [4, 8],
               [5, 7],
               [6, 6],
               [0, 5],
               [1, 4],
               [2, 3]]), tensor([[[ 3, 11],
               [4, 10],
               [5, 9],
               [6,8],
```

```
[0, 7],
         [1, 6],
         [ 2, 5]]]), tensor([[[ 3, 12],
         [4, 11],
        [5, 10],
         [6, 9],
         [0, 8],
         [1, 7],
         [2, 6]]])]
[[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.09090909090909091, 1.0,
0.90909090909091], [0.0, 1.0, 1.0]] [tensor([[[3, 9],
         [4, 8],
         [5, 7],
         [6, 6],
         [0, 5],
         [1, 4],
         [2, 3]]]), tensor([[[ 3, 10],
         [4, 9],
        [5, 8],
         [6, 7],
         [0, 6],
         [1, 5],
         [2, 4]]]), tensor([[[3, 11],
         [4, 10],
         [5,
              9],
         [6,
              8],
         [0, 7],
         [1, 6],
         [ 2, 5]]]), tensor([[[ 3, 12],
         [4, 11],
         [5, 10],
         [6, 9],
         [0, 8],
         [1, 7],
         [2, 6]]])]
[[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]]
[tensor([[[6, 7],
         [0, 6],
         [1, 5],
         [2, 4],
         [3, 3],
         [4, 2],
         [5, 1]]]), tensor([[[6, 8],
         [0, 7],
         [1, 6],
         [2, 5],
         [3, 4],
         [4, 3],
```

```
[5, 2]]]), tensor([[[6, 9],
               [0, 8],
               [1, 7],
               [2, 6],
               [3, 5],
               [4, 4],
               [5, 3]]), tensor([[[ 6, 11],
               [0, 10],
               [1, 9],
               [2, 8],
               [3, 7],
               [4, 6],
               [5, 5]])]
[149]: data=[[[0.7, 1.0, 0.6], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]], [[0.
        40, 1.0, 1.0], [0.0, 1.0, 1.0], [0.09090909090909091, 1.0, 0.
        9090909090909091],[0.0, 1.0, 1.0]],[[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0]
        41.0, 1.0, 1.0, 1.0, 1.0, 1.0
      data
[149]: [[[0.7, 1.0, 0.6], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]],
        [[0.0, 1.0, 1.0],
        [0.0, 1.0, 1.0],
        [0.09090909090909091, 1.0, 0.9090909090909091],
        [0.0, 1.0, 1.0],
        [[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]]]
[43]: import matplotlib.pyplot as plt
      import numpy as np
      data = [
           [[0.7, 1.0, 0.6], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]],
           [[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.09090909090909091, 1.0, 0.
       →9090909090909091], [0.0, 1.0, 1.0]],
           [[0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0], [0.0, 1.0, 1.0]]
      ]
      fig, axs = plt.subplots(nrows=3, ncols=3, figsize=(20,15),)
      x_{labels} = [['2 day', '3 day', '5 day', '6 day'], ['3 day', '4 day', '6 day'], ['3 day', '4 day']
       metrics = ['Mre', 'Ma', 'Mr']
          'hls'
      colors =sns.color_palette("rocket")
      # fig.suptitle('Performance Metrics Across Data Sets', fontsize=2)
      for i, metric in enumerate(metrics):
          for j in range(3):
```

```
values = [x[i] for x in data[j]]
        axs[j, i].bar(np.arange(len(values)), values, __
 ⇔align='center',color=colors,)
        axs[j, i].set_xticks(np.arange(len(values)))
        axs[j, i].set_xticklabels([f'DS {x+1}' for x in range(len(values))],__

fontsize=35)
        axs[j, i].set_ylim([0, 1.2])
        axs[j, i].set_title(f'{metric}, DS_{j+1}', fontsize=35)
        axs[j, i].set_ylabel("")
        axs[j, i].grid(axis='y', linestyle='--')
        for idx, val in enumerate(values):
            axs[j, i].text(idx, val+0.05, f'{val:.1f}', fontsize=30, __
 ⇔ha='center', va='bottom')
        if i == 0:
            axs[j, i].set_xticklabels(x_labels[j], fontsize=35)
              axs[j, i].set_ylabel(metric, fontsize=20)
        else:
            axs[j, i].set_xticks([])
            axs[j, i].set_yticklabels([])
plt.tight_layout()
plt.savefig('pdf\\attn2.pdf', dpi=600)
plt.show()
```



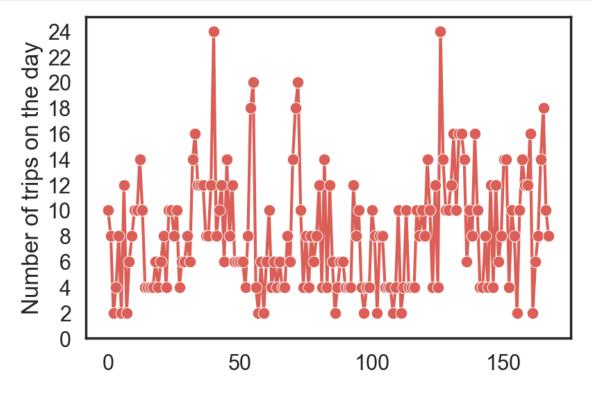
2.1 , ,

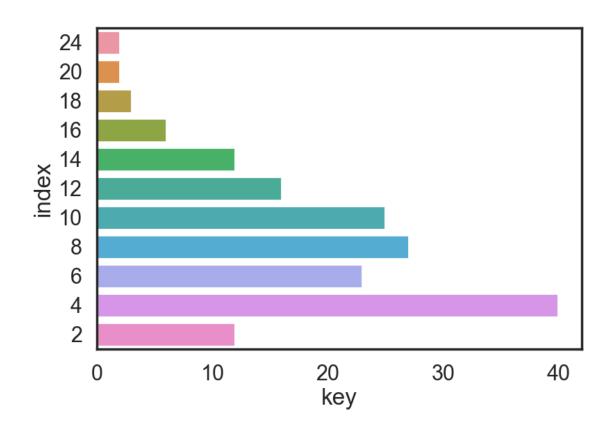
```
label+=[pairs[week_rank][7].split(',')]
   for day in range(7):
       numpair+=[pairs[week_rank][day].split('Attr')[0].split(',')[:-1]]
       traj_day_len.append(len(pairs[week_rank][day].split('Attr')[0].
 ⇔split(',')[:-1]))
       attr.append([[pairs[week_rank][day].split('Attr')[1].split(',')[1:]]])
nested_list = [i[:7] for i in pairs[::49]]
⇔isinstance(sublist, list) else [sublist])]
my_list = flat_list
new_list = []
for sub_list in my_list:
   sub_list = sub_list.split(",")[:-3]
   sub_list = [x for x in sub_list if x != "Attr"]
   new_list.append(",".join(sub_list))
new_dict = {}
for index, item in enumerate(new_list):
   new_dict[index] = item.split(",")
data =new_dict
df = pd.DataFrame(columns=['key','value'])
for key in data:
   for item in data[key]:
       df = df.append({'key':key,'value':item},ignore_index=True)
df[['lon','lat']] = df['value'].str.split('P', expand=True).astype(int)/100
df.to_excel('output.xlsx', index=False)
df['end_lat'] = df['lat'].shift(-1)
df['end_lon'] = df['lon'].shift(-1)
indexes_to_drop = []
for i in range(1, len(df)):
   # key
   if df.loc[i, 'key'] != df.loc[i-1, 'key']:
       indexes_to_drop.append(i-1)
df = df.drop(indexes_to_drop)
df = df.reset_index(drop=True)
```

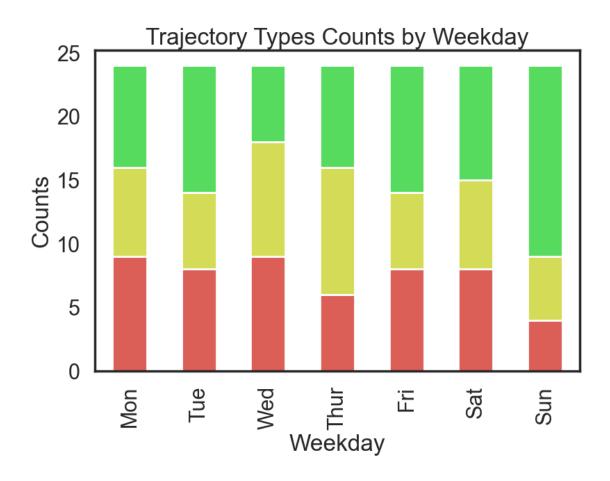
3.0.1

```
[7]: dir="981762.txt"
     # dir[0:7]
     a=f'{dir[0:7]}'
 [7]: '981762.'
[47]: import pandas as pd
     import numpy as np
     import warnings
     warnings.simplefilter(action='ignore', category=FutureWarning)
     dir="981814.txt"
     lines = open('lstm attn\\ \\'+dir, encoding='utf-8').read().strip().
       ⇔split('\n')
     pairs = [[s for s in l.split('#')] for l in lines]
     numpair=[]
                    #
     attr=[]
     traj_day_len=[]#
     label=[]
     for week_rank in range(len(pairs)):
         label+=[pairs[week_rank][7].split(',')]
         for day in range(7):
             numpair+=[pairs[week_rank][day].split('Attr')[0].split(',')[:-1]]
             traj_day_len.append(len(pairs[week_rank][day].split('Attr')[0].
      ⇔split(',')[:-1]))
             attr.append([[pairs[week_rank][day].split('Attr')[1].split(',')[1:]]])
     nested_list = [i[:7] for i in pairs[::49]]
     ⇔isinstance(sublist, list) else [sublist])]
     my list = flat list
     new_list = []
     for sub_list in my_list:
         sub_list = sub_list.split(",")[:-3]
         sub_list = [x for x in sub_list if x != "Attr"]
         new_list.append(",".join(sub_list))
     new_dict = {}
     for index, item in enumerate(new_list):
         new_dict[index] = item.split(",")
     data =new_dict
     df = pd.DataFrame(columns=['key','value'])
```

```
for key in data:
    for item in data[key]:
        df = df.append({'key':key,'value':item},ignore_index=True)
df[['lon', 'lat']] = df['value'].str.split('P', expand=True).astype(int)/100
df.to_excel('output.xlsx', index=False)
df['end_lat'] = df['lat'].shift(-1)
df['end lon'] = df['lon'].shift(-1)
date_counts = df['key'].value_counts()
date counts.sort index(inplace=True)
sns.set context("poster")
sns.set_palette('hls')
plt.figure(figsize=(9, 6))
sns.lineplot(data=date_counts, marker='o', legend=True)
plt.ylabel('Number of trips on the day')
plt.yticks(np.arange(0, max(date_counts)+1, 2))
plt.savefig(f'pdf\{dir[0:7]}_a.pdf', dpi=600)
plt.show()
count_len_traj=pd.DataFrame(date_counts.value_counts()).reset_index().
 →sort_values('index',ascending=False)
plt.figure(figsize=(9, 6))
sns.set_context("poster")
sns.barplot(x='key', y='index',__
 →data=count_len_traj, order=count_len_traj['index'].values, orient='h')
plt.savefig(f'pdf\{dir[0:7]}_b.pdf', dpi=600)
key_counts = df.groupby('key').size().reset_index(name='count')
key_counts['type'] = ''
key counts.loc[(key counts['count'] > 0) & (key counts['count'] < 6), 'type'] = 1
key_counts.loc[(key_counts['count']>5) & (key_counts['count']<10), 'type'] = ___
 key_counts.loc[(key_counts['count']>9), 'type'] = 'chang'
# key_counts week
key_counts['week'] = key_counts['key'] % 7
# week
key_counts['week'] = key_counts['week'].replace({0:'Tue',1:'Wed',2:'Thur',3:
 type_counts = pd.crosstab(key_counts['week'], key_counts['type'])
type_counts = type_counts.reindex(['Mon','Tue','Wed','Thur','Fri','Sat','Sun'])
```







```
[49]:
                               lon
                                      lat end_lat
                                                    end_lon
           key
                     value
                                             30.62
     0
             0 11896P3063
                            118.96
                                    30.63
                                                     118.97
     1
             0 11897P3062 118.97
                                    30.62
                                             30.62
                                                     118.97
     2
                11897P3062 118.97
                                    30.62
                                             30.63
                                                     118.98
```

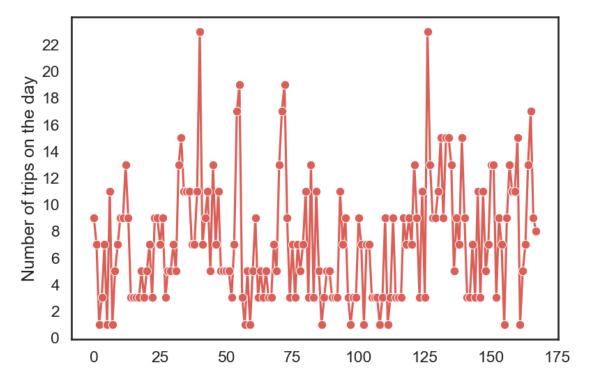
```
3
           11898P3063
                                30.63
                                          30.63
                                                   118.98
        0
                        118.98
4
        0
           11898P3063
                        118.98
                                30.63
                                          30.62
                                                   118.97
                                          31.32
                                                   121.45
      167
           12145P3132
                        121.45
                                31.32
1214
1215
      167
           12145P3132
                        121.45
                                31.32
                                          31.33
                                                   121.44
1216
           12144P3133
                        121.44
                                31.33
                                          31.33
                                                   121.44
      167
1217
                                          31.33
                                                   121.43
      167
           12144P3133
                        121.44
                                31.33
1218
      167
           12143P3133
                        121.43
                                31.33
                                            NaN
                                                      NaN
```

[1219 rows x 6 columns]

```
[50]: date_counts = df['key'].value_counts()

#

date_counts.sort_index(inplace=True)
sns.set_palette('hls')
plt.figure(figsize=(9, 6))
sns.lineplot(data=date_counts, marker='o', legend=True)
plt.ylabel('Number of trips on the day')
plt.yticks(np.arange(0, max(date_counts)+1, 2))
plt.savefig('eps\summ_33.eps', dpi=600, format='eps')
plt.show()
```



```
[51]: count_len_traj=pd.DataFrame(date_counts.value_counts()).reset_index().

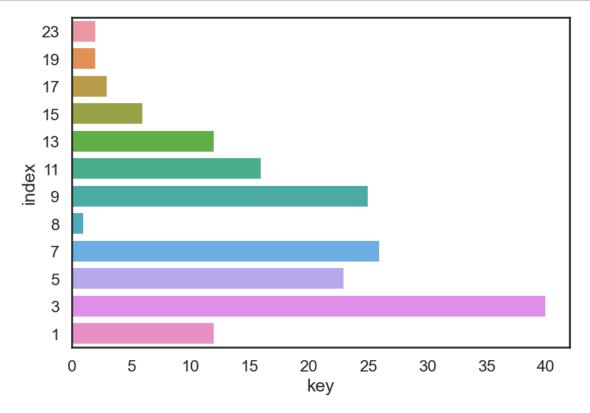
sort_values('index',ascending=False)

plt.figure(figsize=(9, 6))

sns.barplot(x='key', y='index',__

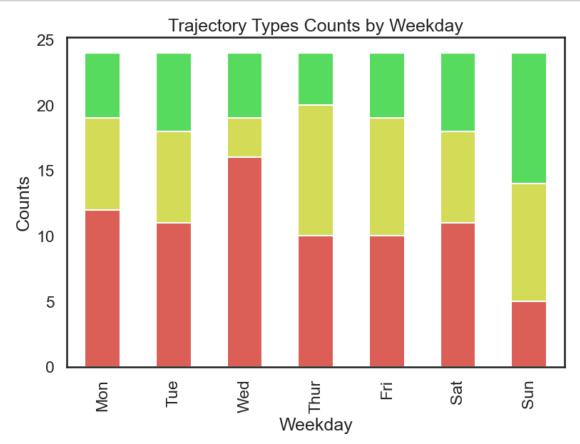
data=count_len_traj,order=count_len_traj['index'].values, orient='h')

plt.savefig('eps\summ_34.eps', dpi=600, format='eps')
```



```
[52]: # key
key_counts = df.groupby('key').size().reset_index(name='count')
key_counts['type'] = ''
key_counts.loc[(key_counts['count']>0) & (key_counts['count']<6), 'type'] = \( \text{"duan'} \)
key_counts.loc[(key_counts['count']>5) & (key_counts['count']<10), 'type'] = \( \text{"zhong'} \)
key_counts.loc[(key_counts['count']>9), 'type'] = 'chang'

# key_counts week
key_counts['week'] = key_counts['key'] % 7
# week
key_counts['week'] = key_counts['week'].replace({0:'Tue',1:'Wed',2:'Thur',3:
    \( \text{"Fri',4:'Sat',5:'Sun',6:'Mon'}) \)
#
```



3.1

```
[102]: import pandas as pd
       import numpy as np
       dir="981814.txt"
       lines = open('lstm attn\\ \\'+dir, encoding='utf-8').read().strip().
        ⇔split('\n')
       pairs = [[s for s in l.split('#')] for l in lines]
       numpair=[]
       attr=[]
       traj_day_len=[]#
       label=[]
       for week rank in range(len(pairs)):
           label+=[pairs[week_rank][7].split(',')]
           for day in range(7):
               numpair+=[pairs[week_rank][day].split('Attr')[0].split(',')[:-1]]
               traj_day_len.append(len(pairs[week_rank][day].split('Attr')[0].

¬split(',')[:-1]))
               attr.append([[pairs[week_rank][day].split('Attr')[1].split(',')[1:]]])
       nested_list = [i[:7] for i in pairs[::49]]
       flat list = [item for sublist in nested list for item in (sublist if,
        sisinstance(sublist, list) else [sublist])]
       my_list = flat_list
       new_list = []
       for sub_list in my_list:
           sub_list = sub_list.split(",")[:-3]
           sub_list = [x for x in sub_list if x != "Attr"]
           new_list.append(",".join(sub_list))
       new_dict = {}
       for index, item in enumerate(new_list):
           new_dict[index] = item.split(",")
       data =new_dict
       df = pd.DataFrame(columns=['key','value'])
       for key in data:
           for item in data[key]:
               df = df.append({'key':key,'value':item},ignore_index=True)
       df[['lon','lat']] = df['value'].str.split('P', expand=True).astype(int)/100
       df.to_excel('output.xlsx', index=False)
       df['end_lat'] = df['lat'].shift(-1)
       df['end_lon'] = df['lon'].shift(-1)
       df = df[(df['lat'] != df['end_lat']) | (df['lon'] != df['end_lon'])]
```

```
[103]: import pydeck as pdk
       import pandas as pd
       layer = pdk.Layer(
           "ScreenGridLayer",
           df,
           pickable=False,
           opacity=0.8,
           cell_size_pixels=60,
             color_range=[
                 [0, 25, 0, 25],
       #
                 [0, 85, 0, 85],
                 [0, 127, 0, 127],
       #
                 [0, 170, 0, 170],
                 [0, 190, 0, 190],
                 [0, 255, 0, 255],
       #
             ],#
           get_position="COORDINATES",
           get_weight="SPACES",
       # Set the viewport location
       view_state = pdk.ViewState(latitude=df.iloc[0,0][1], longitude=df.iloc[0,0][0],__
        ⇒zoom=12, bearing=0, pitch=0)
       # Render
       r = pdk.
        Geck(layers=[layer],initial_view_state=view_state,map_style='light_no_labels')#
       r.to_html("eps\summ_32.html")
```

[103]: <IPython.core.display.HTML object>

4

4.1 1

```
[30]: import pandas as pd
    df=pd.read_excel("6 3 .xlsx",sheet_name=['data1','data2'])['data2']
    raw_data=df.melt(['model','attr','DS'],['MRE','MA','MR'],'index','value')
    raw=raw_data[raw_data['attr']=='mean']
    raw
```

```
[30]:
                model
                        attr
                                 DS index
                                               value
                              DS 1
      5
                   ED
                        mean
                                      MRE
                                            0.137000
      6
                 TAED
                        mean
                              DS 1
                                      MRE
                                            0.075000
      7
                              DS 1
                HTAED
                        mean
                                      MRE
                                            0.059000
      8
            HTAED-GRU
                              DS_1
                                      MRE
                        mean
                                            0.092806
      9
                              DS_1
              HTIA-UP
                        mean
                                      MRE
                                            0.040763
      20
                              DS_2
                                      MRE
                   ED
                                            0.303000
                        mean
      21
                              DS_2
                                      MRE
                 TAED
                        mean
                                            0.236000
      22
                               DS_2
                HTAED
                                      MRE
                                            0.161000
                        mean
      23
                               DS_2
            HTAED-GRU
                                      MRE
                        mean
                                            0.118348
      24
                               DS_2
                                      MRE
              HTIA-UP
                        mean
                                            0.052700
      35
                              DS_3
                                      MRE
                                            0.207000
                   ED
                        mean
      36
                 TAED
                              DS_3
                                      MRE
                                            0.169000
                        mean
      37
                              DS_3
                HTAED
                                      MRE
                                            0.143000
                        mean
      38
            HTAED-GRU
                              DS_3
                                      MRE
                                            0.112051
                        mean
                              DS 3
      39
              HTIA-UP
                        mean
                                      MRE
                                            0.068234
                               DS_1
      46
                 HTIA
                        mean
                                      MRE
                                            0.051323
      49
                 HTIA
                               DS 2
                                      MRE
                        mean
                                            0.078699
                 HTIA
                              DS_3
      52
                        mean
                                      MRE
                                            0.041432
      59
                   ED
                              DS_1
                                       MA
                                            0.911000
                        mean
      60
                 TAED
                        mean
                              DS_1
                                       MA
                                            0.953000
      61
                HTAED
                              DS_1
                                            0.965000
                        mean
                                       MA
      62
            HTAED-GRU
                               DS_1
                        mean
                                       MA
                                            0.981757
      63
              HTIA-UP
                               DS_1
                        mean
                                       MA
                                            0.990884
      74
                               DS 2
                   ED
                        mean
                                        MA
                                            0.792000
      75
                 TAED
                        mean
                              DS_2
                                       MA
                                            0.839000
      76
                HTAED
                              DS_2
                                       MA
                                            0.883000
                        mean
      77
            HTAED-GRU
                              DS_2
                                       MA
                                            0.981560
                        mean
      78
              HTIA-UP
                        mean
                              DS_2
                                       MA
                                            0.992982
                              DS_3
      89
                   ED
                        mean
                                       MA
                                            0.856000
      90
                              DS 3
                 TAED
                                       MA
                                            0.877000
                        mean
                               DS 3
      91
                HTAED
                        mean
                                       MA
                                            0.895000
      92
            HTAED-GRU
                               DS_3
                        mean
                                       MA
                                            0.985876
      93
              HTIA-UP
                        mean
                              DS_3
                                            0.989886
      100
                 HTIA
                              DS_1
                                       MA
                        mean
                                            0.990302
                              DS_2
      103
                 HTIA
                        mean
                                       MΑ
                                            0.981416
      106
                 HTIA
                              DS_3
                                       MA
                                            0.995201
                        mean
      113
                   ED
                              DS_1
                                       MR
                                            0.889000
                        mean
```

```
DS_1
      115
              HTAED
                                   MR 0.956000
                     mean
      116 HTAED-GRU
                     mean
                           DS 1
                                   MR 0.969419
      117
            HTIA-UP
                     mean
                           DS_1
                                   MR 0.988327
      128
                     mean DS_2
                 ED
                                   MR 0.794000
      129
               TAED
                     mean DS_2
                                   MR 0.840000
      130
              HTAED
                     mean DS 2
                                   MR 0.874000
      131 HTAED-GRU
                     mean DS 2
                                   MR 0.965682
      132
                     mean DS 2
            HTIA-UP
                                   MR 0.975431
      143
                     mean DS 3
                                   MR 0.846000
                 ED
                     mean DS 3
      144
               TAED
                                   MR 0.871000
      145
              HTAED
                     mean DS_3
                                   MR 0.894000
      146 HTAED-GRU
                     mean DS 3
                                   MR 0.960650
      147
            HTIA-UP
                     mean
                           DS_3
                                   MR 0.976155
      154
               HTIA
                     mean DS_1
                                   MR 0.979661
      157
               HTIA
                     mean DS_2
                                   MR 0.972718
      160
               HTIA mean DS_3
                                   MR 0.984073
[31]: import seaborn as sns
      import matplotlib.pyplot as plt
      sns.set theme(style="white", context="poster")
      # Set up the matplotlib figure
      f, (ax1, ax2, ax3) = plt.subplots(3, 1, figsize=(9, 6), sharex=True)
      sns.barplot(data=raw.query('index=="MRE" '),x='DS',y='value',hue='model',ax=ax1)
      sns.barplot(data=raw[raw['index']=='MA'],x='DS',y='value',hue='model',ax=ax2)
      ax2.set_ylim([0.78, 1])
      sns.barplot(data=raw[raw['index']=='MR'],x='DS',y='value',hue='model',ax=ax3)
      ax3.set_ylim([0.78, 1])
      ax1.legend_.remove()
      ax2.legend_.remove()
      ax3.legend .remove()
      ax1.set_xlabel('')
      ax2.set xlabel('')
      ax1.set_ylabel('Mre')
      ax2.set ylabel('Ma')
      ax3.set ylabel('Mr')
      plt.legend(title='Model', loc='upper center', bbox to anchor=(0.5,4.8), ncol=3)
      plt.xlabel('Dataset')
      sns.despine(bottom=True)
```

MR 0.942000

114

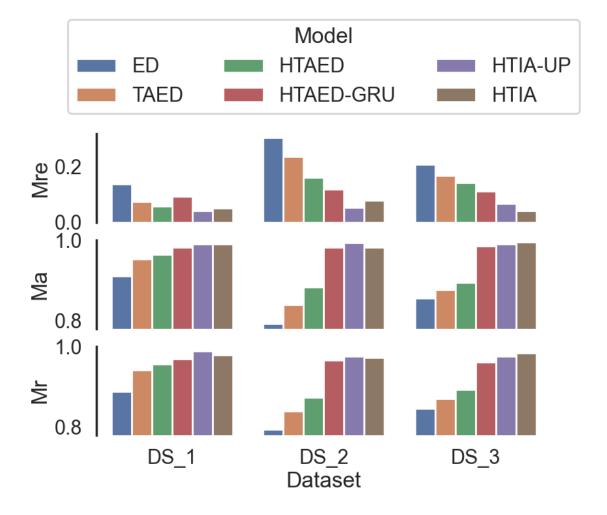
TAED

mean

 DS_1

plt.savefig('pdf\\result1.pdf', dpi=300,bbox_inches = 'tight')

plt.setp(f.axes, yticks=[])
plt.tight_layout(h_pad=2)



4.2

```
[128]: import pandas as pd
df=pd.read_excel("6 3 .xlsx",sheet_name=['data1','data2'])['data1']
raw_data=df.melt(['model','attr','DS'],['MRE','MA','MR'],'index','value')
raw=raw_data[raw_data['attr']=='mean']
raw
```

```
[128]:
                   model
                           attr
                                   DS index
                                                 value
       3
                    HTIA
                                DS_1
                                        MRE
                                             0.051323
                           mean
                 HTIA-UP
       4
                           mean
                                DS_1
                                        MRE
                                             0.040763
       5
           T-Transformer
                           mean
                                 DS_1
                                        MRE
                                              0.069890
       12
                    HTIA
                           mean DS_2
                                        MRE
                                              0.078699
       13
                 HTIA-UP
                                 DS_2
                                        MRE
                                             0.052700
                           mean
       14
                                 DS 2
           T-Transformer
                                        MRE
                                             0.035963
                           mean
                                 DS_3
       21
                                        MRE
                    HTIA
                           mean
                                              0.041432
       22
                 HTIA-UP
                          mean
                                 DS_3
                                        MRE
                                             0.068234
```

```
30
                   HTIA
                         mean DS_1
                                       MA 0.990302
                         mean DS_1
      31
                HTIA-UP
                                       MA 0.990884
      32
          T-Transformer
                         mean DS_1
                                       MA 0.982825
      39
                         mean DS_2
                   HTIA
                                       MA 0.981416
      40
                HTIA-UP
                         mean DS_2
                                       MA 0.992982
          T-Transformer mean DS 2
      41
                                       MA 0.994026
                   HTIA mean DS_3
      48
                                       MA 0.995201
                                       MA 0.989886
      49
                HTIA-UP
                         mean DS 3
          T-Transformer mean DS 3
      50
                                       MA 0.975459
                                       MR 0.979661
      57
                   HTIA mean DS 1
      58
                HTIA-UP
                         mean DS_1
                                       MR 0.988327
      59
          T-Transformer mean DS 1
                                       MR 0.975898
      66
                   HTIA
                         mean DS_2
                                       MR 0.972718
      67
                HTIA-UP
                         mean DS 2
                                       MR 0.975431
      68
          T-Transformer
                         mean DS_2
                                       MR 0.993379
      75
                         mean DS_3
                                       MR 0.984073
                   HTIA
      76
                                       MR 0.976155
                HTIA-UP
                         mean DS 3
          T-Transformer mean DS_3
                                       MR 0.982795
[129]: import seaborn as sns
      import matplotlib.pyplot as plt
      sns.set_theme(style="white", context="paper")
       # Set up the matplotlib figure
      f, (ax1, ax2, ax3) = plt.subplots(3, 1, figsize=(4.2,5), sharex=True)
      hue_order=['T-Transformer', 'HTIA', 'HTIA-UP']
      sns.barplot(data=raw.query('index=="MRE"__
       o'),x='DS',y='value',hue='model',hue_order=hue_order,ax=ax1)
        abarplot(data=raw[raw['index']=='MA'],x='DS',y='value',hue='model',hue_order=hue_order,ax=ax
      ax2.set_ylim([0.95, 1])
        abarplot(data=raw[raw['index']=='MR'],x='DS',y='value',hue='model',hue_order=hue_order,ax=ax
      ax3.set ylim([0.95, 1])
      ax1.legend_.remove()
      ax2.legend_.remove()
      ax3.legend_.remove()
      ax1.set_xlabel('')
      ax2.set_xlabel('')
      ax1.set_ylabel('Mre')
      ax2.set_ylabel('Ma')
      ax3.set_ylabel('Mr')
      plt.legend(title='Model', loc='lower center', bbox_to_anchor=(0.5, 3.32),__
      plt.xlabel('DataSet')
      sns.despine(bottom=True)
```

MRE 0.073580

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T-Transformer mean DS_3

```
# plt.setp(f.axes, yticks=[])
# plt.tight_layout(h_pad=2)
plt.savefig('eps\\result2.eps', dpi=600, format='eps')
```

The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.

