

# prefix\_identification\_noELBO

September 24, 2019

```
In [109]: import numpy as np
          from scipy.spatial.distance import euclidean
          import scipy
          import torch

          import dtw
          import dtwalign
          from fastdtw import fastdtw

          from sklearn.manifold import TSNE
          import pandas as pd

          from src.speech_classification.audio_processing import AudioPreprocessorFbank, Speech
          from src.siamese_net_sound_similarity.slstm_train import SiameseSpeechCommandsDataCo
          from src.siamese_net_sound_similarity.train import SiameseLSTMNet

          import matplotlib.pyplot as plt
          import matplotlib
          import seaborn as sns
          sns.set()

          %matplotlib inline
```

## 0.1 Load data

```
In [110]: wanted_words = ['bed', 'bird', 'cat', 'dog', 'down', 'eight', 'five', 'four', 'go',
                           'marvin',
                           'nine', 'no', 'off', 'on', 'one', 'right', 'seven', 'sheila', 'six',
                           'up', 'wow', 'yes', 'zero']

          wanted_words_combined = wanted_words

          model_settings = {
              'dct_coefficient_count': 26,
              'label_count': len(wanted_words_combined) + 2,
              'hidden_reccurent_cells_count': 128,
```

```

        'winlen': 0.04,
        'winstep': 0.02
    }

In [111]: preproc = AudioPreprocessorFbank(nfilt=model_settings['dct_coefficient_count'], winlen=winlen,
                                           winstep=model_settings['winstep'])

data_iter = SiameseSpeechCommandsDataCollector(preproc,
                                                data_dir=r'C:\Study\Speech_command_classification',
                                                wanted_words=wanted_words_combined,
                                                testing_percentage=10,
                                                validation_percentage=10
                                                )

index_to_word = {v:k for k,v in data_iter.word_to_index.items()}

In [112]: n_mini_batch_size = 1
data = data_iter.get_data(n_mini_batch_size, 0, 'training')
labels = data['y']

duplicates = data_iter.get_duplicates(labels, 0, 'training')
assert np.any(labels == duplicates['y'])

non_duplicates = data_iter.get_nonduplicates(labels, 0, 'training')
assert np.any(labels != non_duplicates['y'])

In [113]: i = np.random.randint(data['y'].shape[0])
sample_idx = i

fig, axes = plt.subplots(2, 3, figsize=(15,10))

# plt.figure()
im = axes[0][0].imshow(data['x'][i].T)
# fig.colorbar(im)
label = index_to_word[data['y'][0]]
axes[0][0].title.set_text(f'{label}')

axes[0][1].imshow(duplicates['x'][i].T)
# plt.colorbar()
label = index_to_word[data['y'][0]]
axes[0][1].title.set_text(f'{label}')

axes[0][2].imshow(non_duplicates['x'][i].T)
# plt.colorbar()
label = index_to_word[non_duplicates['y'][0]]
axes[0][2].title.set_text(f'{label}')

```

```
##### DTW

x = data['x'][i]

# self dtw (expected to be 0)

y = data['x'][i]
res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
res.plot_path(ax=axes[1][0])
axes[1][0].set_title(f"anchor DTW = {res.normalized_distance:.4f}")

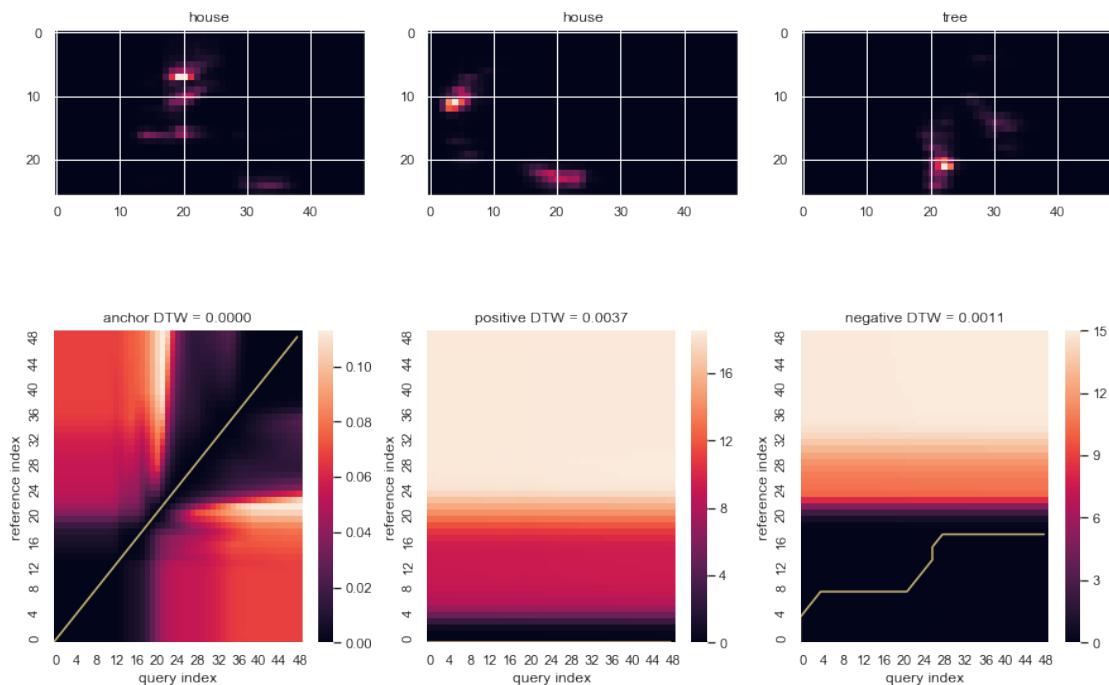
# positive dtw

y = duplicates['x'][sample_idx]
res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
res.plot_path(ax=axes[1][1])
axes[1][1].set_title(f"positive DTW = {res.normalized_distance:.4f}")

# negative dtw

y = non_duplicates['x'][sample_idx]
res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
res.plot_path(ax=axes[1][2])
axes[1][2].set_title(f"negative DTW = {res.normalized_distance:.4f}")
```

Out[113]: Text(0.5,1,'negative DTW = 0.0011')



# 1 Gather Data for analysis

```
In [172]: n_mini_batch_size = 1000
```

```
# gather data
```

```
data = data_iter.get_data(n_mini_batch_size, 0, 'training')
labels = data['y']
```

```
duplicates = data_iter.get_duplicates(labels, 0, 'training')
assert np.any(labels == duplicates['y'])
```

```
non_duplicates = data_iter.get_nonduplicates(labels, 0, 'training')
assert np.any(labels != non_duplicates['y'])
```

## 1.1 DTW distribution on raw data

```
In [173]: # initialize dist_lists
```

```
duplicates_dtw = []
non_duplicates_dtw = []
```

```
for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = data['x'][i].squeeze()
```

```
    y_duplicate = duplicates['x'][i].squeeze()
    duplicates_dtw.append(dtwalign.dtw(x, y_duplicate, dist=euclidean, open_end=False))
```

```
    y_non_duplicate = non_duplicates['x'][i].squeeze()
    non_duplicates_dtw.append(dtwalign.dtw(x, y_non_duplicate, dist=euclidean, open_end=False))
```

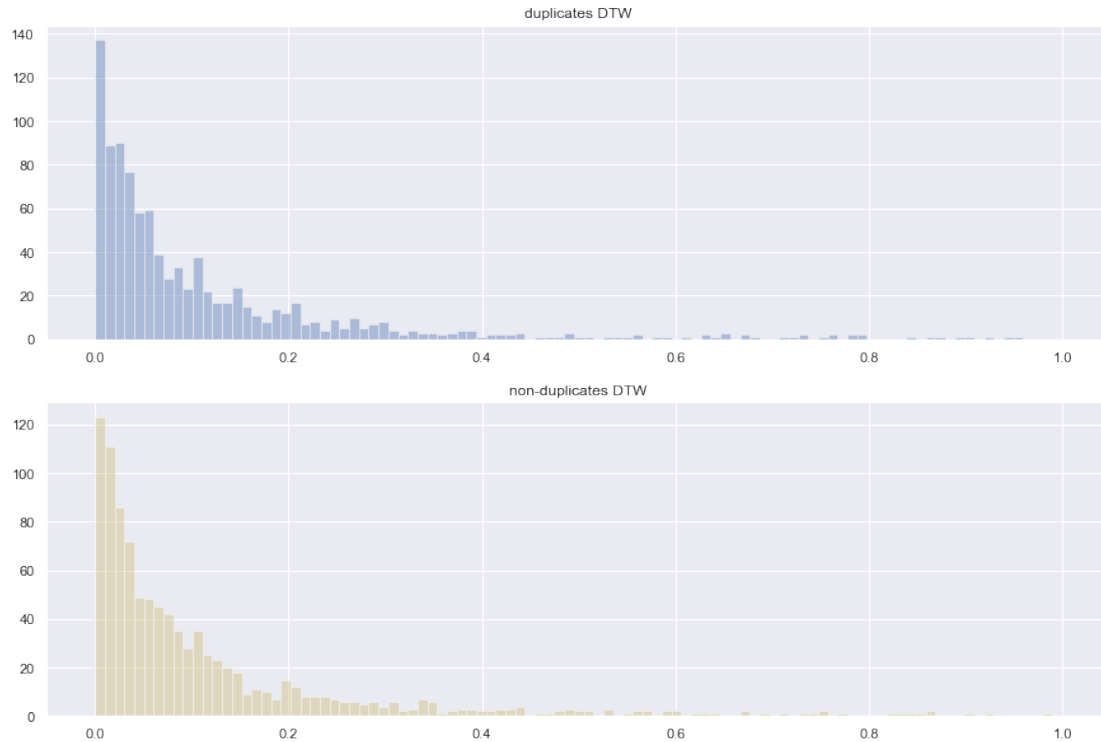
999

```
In [174]: bins = np.linspace(0, 1., 100)
fig, axes = plt.subplots(2, 1, figsize=(15,10))
```

```
duplicates_dtw_filtered = np.array(duplicates_dtw)[np.where(np.array(duplicates_dtw) > 0)]
non_duplicates_dtw_filtered = np.array(non_duplicates_dtw)[np.where(np.array(non_duplicates_dtw) > 0)]
```

```
sns.distplot(duplicates_dtw_filtered, bins=bins, ax=axes[0], kde=False)
axes[0].title.set_text('duplicates DTW')
```

```
sns.distplot(non_duplicates_dtw_filtered, bins=bins, color='y', ax=axes[1], kde=False)
axes[1].title.set_text('non-duplicates DTW')
plt.show()
```



### 1.1.1 KL Divergence between two DTW distributions:

```
In [175]: hist_dup, _ = np.histogram(duplicates_dtw, bins=bins)
          hist_dup = hist_dup / np.sum(hist_dup)

          hist_non_dup, _ = np.histogram(non_duplicates_dtw, bins=bins)
          hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

          dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
          print(f"DKL is: {dkl:.4f}")
```

DKL is: 0.1177

1.1.2 As it can be seen, DTW calculated on MFCCs signals failed to capture different and similar samples showing no significant difference between duplicates and non-duplicates pairs. Kullback-Leibler divergence just proves this fact that P (distribution of distances between duplicate samples) and Q (non-duplicate ones) have no significant difference.

## 2 Improving DTW using latent variables from LSTM classifier as a signal

```
In [229]: # load nn
```

```
nn_fname = r'C:/Study/SpeechAcquisitionModel/reports/seamise_net_09_23_2019_03_35_PM
nn = torch.load(nn_fname)
```

```
In [230]: nn_input = torch.from_numpy(data['x']).cuda().float()
```

```
In [231]: z, *_ = nn.single_forward(nn_input)
z = z.detach().cpu().numpy()
```

```
duplicates_z, *_ = nn.single_forward(torch.from_numpy(duplicates['x']).cuda().float())
duplicates_z = duplicates_z.detach().cpu().numpy()
```

```
non_duplicates_z, *_ = nn.single_forward(torch.from_numpy(non_duplicates['x']).cuda())
non_duplicates_z = non_duplicates_z.detach().cpu().numpy()
```

### 2.1 Embedding and DTW

```
In [232]: # random latent variables
```

```
# z, _, _, _ = nn.single_forward(nn_input)
# z = z.detach().cpu().numpy()
```

```
# duplicates_z, _, _, _ = nn.single_forward(torch.from_numpy(duplicates['x']).cuda())
# duplicates_z = duplicates_z.detach().cpu().numpy()
```

```
# non_duplicates_z, _, _, _ = nn.single_forward(torch.from_numpy(non_duplicates['x'])
# non_duplicates_z = non_duplicates_z.detach().cpu().numpy()
```

```
In [233]: i = np.random.randint(data['y'].shape[0])
sample_idx = i
```

```
fig, axes = plt.subplots(2, 3, figsize=(15,10))
```

```
# plt.figure()
im = axes[0][0].imshow(z[i].T, vmin=-1, vmax=1)
# fig.colorbar(im)
label = index_to_word[data['y'][i]]
axes[0][0].title.set_text(f'{label}')
```

```

axes[0][1].imshow(duplicates_z[i].T, vmin=-1, vmax=1)
# plt.colorbar()
label = index_to_word[data['y'][i]]
axes[0][1].title.set_text(f'{label}')

axes[0][2].imshow(non_duplicates_z[i].T, vmin=-1, vmax=1)
# plt.colorbar()
label = index_to_word[non_duplicates['y'][i]]
axes[0][2].title.set_text(f'{label}')

##### DTW

x = z[sample_idx]

# self dtw (expected to be 0)

y = z[sample_idx]
res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
res.plot_path(ax=axes[1][0])
axes[1][0].set_title(f"anchor DTW = {res.normalized_distance:.4f}")

# positive dtw

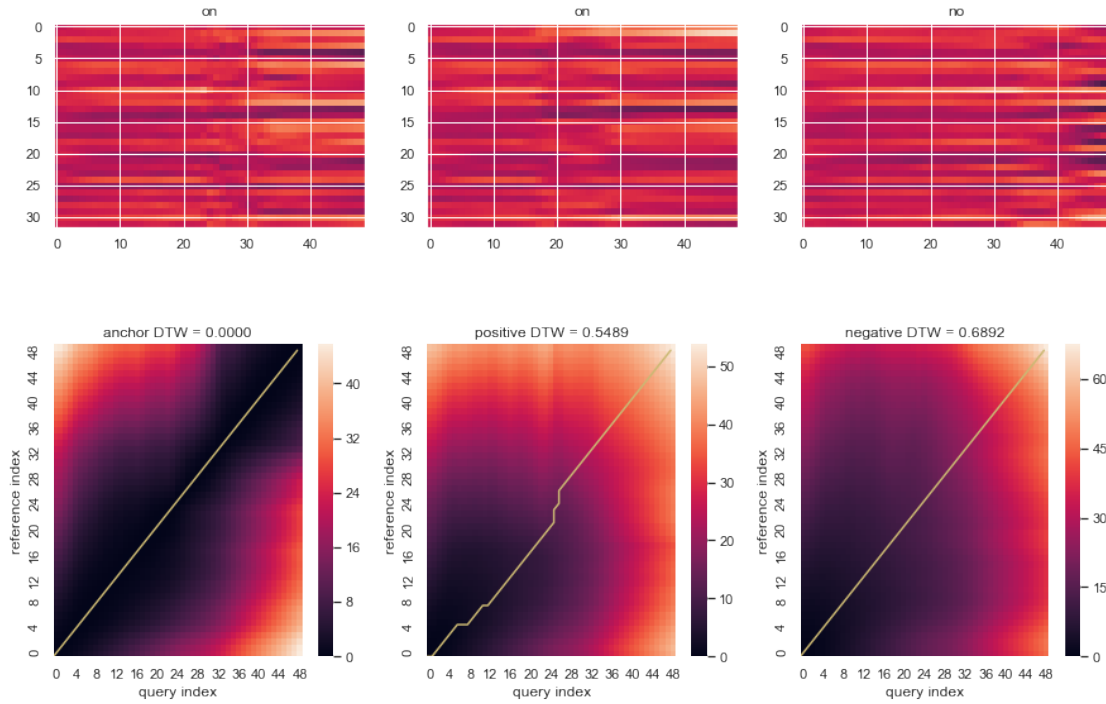
y = duplicates_z[sample_idx]
res = dtwalign.dtw(x, y, dist=euclidean, open_end=False)
res.plot_path(ax=axes[1][1])
axes[1][1].set_title(f"positive DTW = {res.normalized_distance:.4f}")

# negative dtw

y = non_duplicates_z[sample_idx]
res = dtwalign.dtw(x, y, dist=euclidean, open_end=False)
res.plot_path(ax=axes[1][2])
axes[1][2].set_title(f"negative DTW = {res.normalized_distance:.4f}")

```

Out[233]: Text(0.5,1,'negative DTW = 0.6892')



### 3 1.1 DTW distribution in latent space z (open\_end=True)

In [234]: # initialize dist\_lists

```

duplicates_dtw_z = []
non_duplicates_dtw_z = []

for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = z[i].squeeze()

    y_duplicate = duplicates_z[i].squeeze()
    duplicates_dtw_z.append(dtwalign.dtw(x, y_duplicate, dist=euclidean, open_end=True))

    y_non_duplicate = non_duplicates_z[i].squeeze()
    non_duplicates_dtw_z.append(dtwalign.dtw(x, y_non_duplicate, dist=euclidean, open_end=True))

```

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```

In [235]: bins = np.linspace(0, 1., 100)
fig, axes = plt.subplots(1, 1, figsize=(15,8))

sns.distplot(duplicates_dtw_z, ax=axes, bins=bins, label="Positives")

```



```

axes.title.set_text('duplicates DTW')

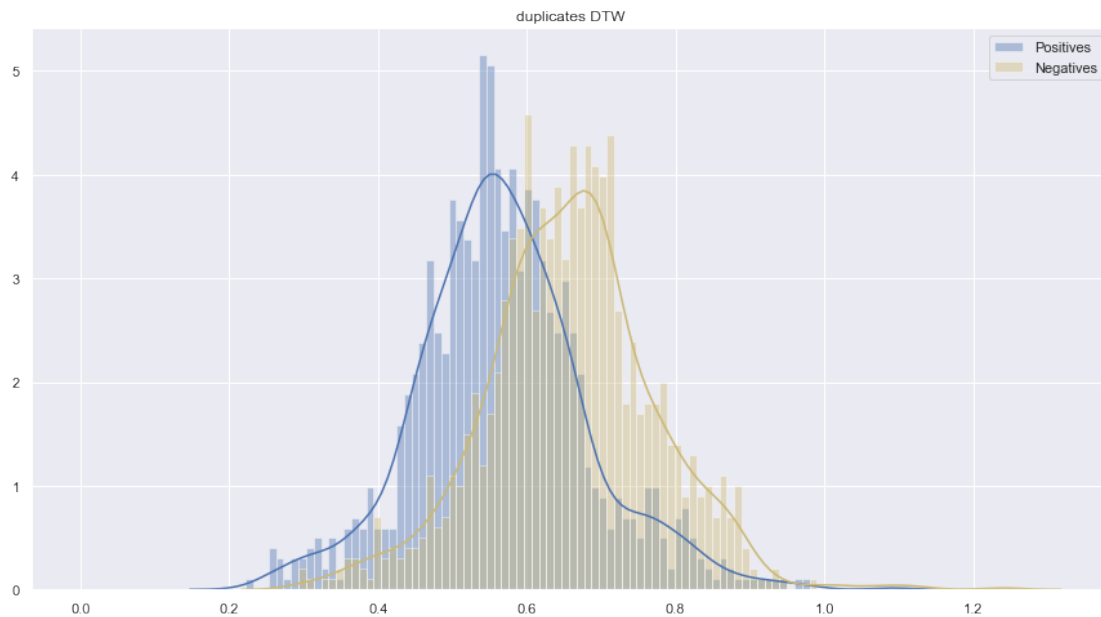
sns.distplot(non_duplicates_dtw_z, bins=bins, color='y', ax=axes, label='Negatives')
axes.legend()
# axes[1].title.set_text('non-duplicates DTW')
plt.show()

hist_dup, _ = np.histogram(duplicates_dtw_z, bins=bins)
hist_dup = hist_dup / np.sum(hist_dup)

hist_non_dup, _ = np.histogram(non_duplicates_dtw_z, bins=bins)
hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
print(f"DKL is: {dkl:.4f}")

```



DKL is: 0.5164

## 4 1.2 DTW distribution in latent space z (open\_end=False)

```

In [236]: # initialize dist_lists

duplicates_dtw_z = []
non_duplicates_dtw_z = []

```

```

for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = z[i].squeeze()

    y_duplicate = duplicates_z[i].squeeze()
    duplicates_dtw_z.append(dtwalign.dtw(x, y_duplicate, dist=euclidean, open_end=False))

    y_non_duplicate = non_duplicates_z[i].squeeze()
    non_duplicates_dtw_z.append(dtwalign.dtw(x, y_non_duplicate, dist=euclidean, open_end=False))

```

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```

In [237]: bins = np.linspace(0, 1.5, 100)
fig, axes = plt.subplots(1, 1, figsize=(15,8))

sns.distplot(duplicates_dtw_z, ax=axes, bins=bins, label="Positives")
axes.title.set_text('duplicates DTW')

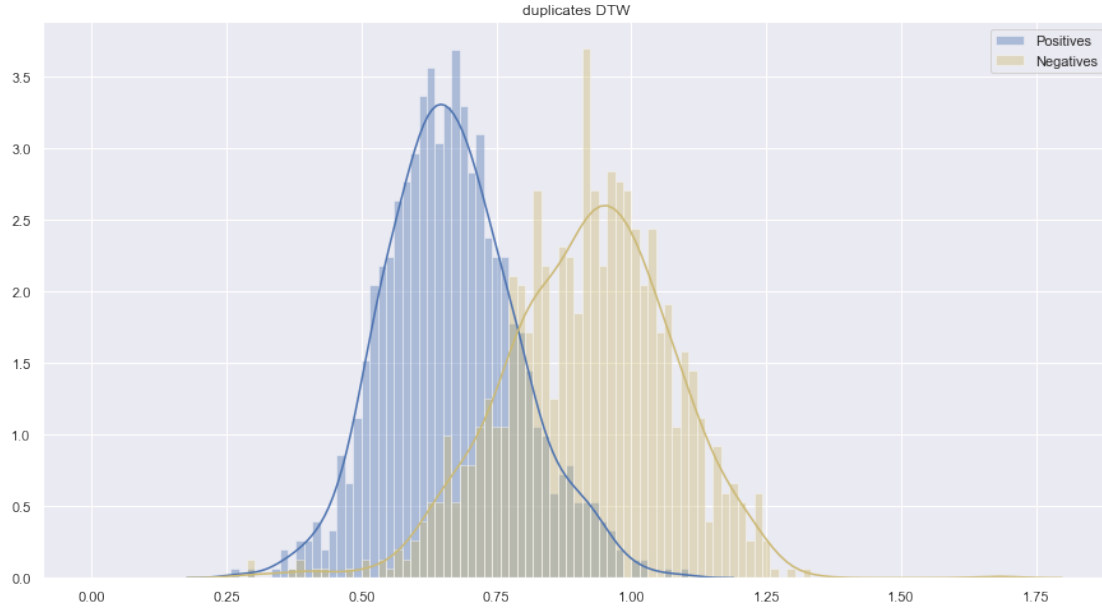
sns.distplot(non_duplicates_dtw_z, bins=bins, color='y', ax=axes, label='Negatives')
axes.legend()
# axes[1].title.set_text('non-duplicates DTW')
plt.show()

hist_dup, _ = np.histogram(duplicates_dtw_z, bins=bins)
hist_dup = hist_dup / np.sum(hist_dup)

hist_non_dup, _ = np.histogram(non_duplicates_dtw_z, bins=bins)
hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
print(f"DKL is: {dkl:.4f}")

```



DKL is: 1.7515

## 5 2. DTW (prefix case $n \ll m$ ) $n = 40$

```
In [238]: PREFIX_LEN = 40
```

```
In [239]: nn_input = torch.from_numpy(data['x'][:, :PREFIX_LEN, :]).cuda().float()
```

```
In [240]: z, *_ = nn.single_forward(nn_input)
z = z.detach().cpu().numpy()
```

```
duplicates_z, *_ = nn.single_forward(torch.from_numpy(duplicates['x']).cuda().float())
duplicates_z = duplicates_z.detach().cpu().numpy()
```

```
non_duplicates_z, *_ = nn.single_forward(torch.from_numpy(non_duplicates['x']).cuda())
non_duplicates_z = non_duplicates_z.detach().cpu().numpy()
```

```
In [241]: # random latent variables
# z, _, _, _ = nn.single_forward(nn_input)
# z = z.detach().cpu().numpy()

# duplicates_z, _, _, _ = nn.single_forward(torch.from_numpy(duplicates['x']).cuda())
# duplicates_z = duplicates_z.detach().cpu().numpy()

# non_duplicates_z, _, _, _ = nn.single_forward(torch.from_numpy(non_duplicates['x']).cuda())
# non_duplicates_z = non_duplicates_z.detach().cpu().numpy()
```

```

In [242]: i = np.random.randint(data['y'].shape[0])
          sample_idx = i

          fig, axes = plt.subplots(2, 3, figsize=(15,10))

          # plt.figure()
          im = axes[0][0].imshow(z[i].T, vmin=-1, vmax=1)
          # fig.colorbar(im)
          label = index_to_word[data['y'][i]]
          axes[0][0].title.set_text(f'{label}')

          axes[0][1].imshow(duplicates_z[i].T, vmin=-1, vmax=1)
          # plt.colorbar()
          label = index_to_word[data['y'][i]]
          axes[0][1].title.set_text(f'{label}')

          axes[0][2].imshow(non_duplicates_z[i].T, vmin=-1, vmax=1)
          # plt.colorbar()
          label = index_to_word[non_duplicates['y'][i]]
          axes[0][2].title.set_text(f'{label}')

          ##### DTW

          x = z[sample_idx]

          # self dtw (expected to be 0)

          y = z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][0])
          axes[1][0].set_title(f"anchor DTW = {res.normalized_distance:.4f}")

          # positive dtw

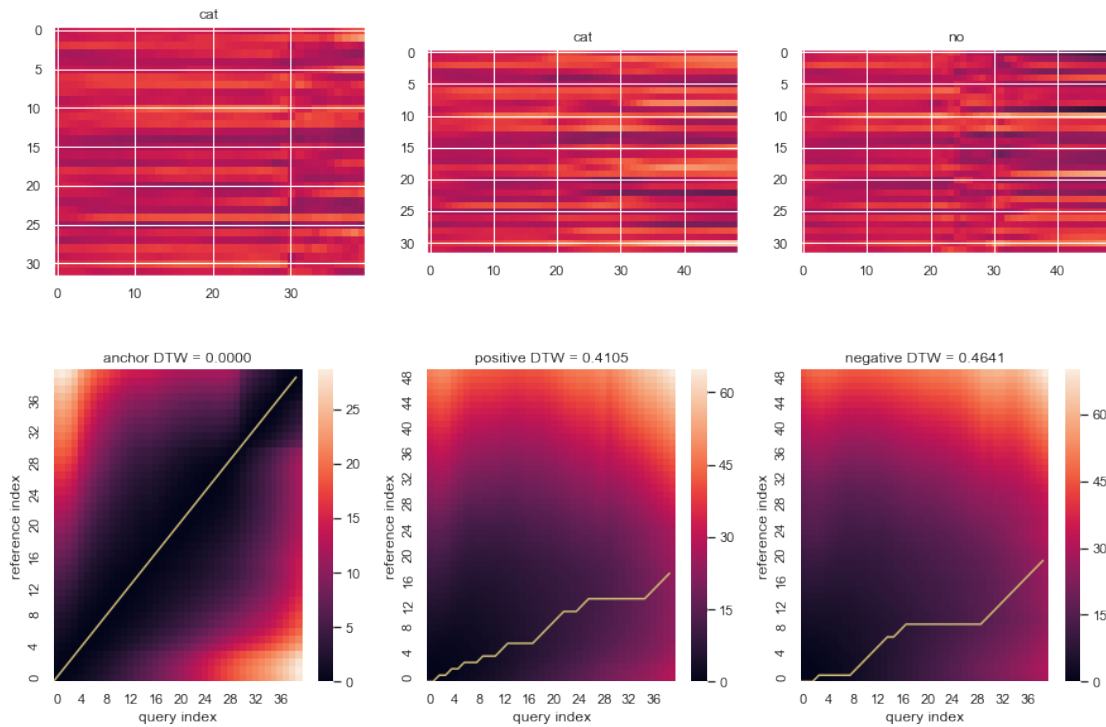
          y = duplicates_z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][1])
          axes[1][1].set_title(f"positive DTW = {res.normalized_distance:.4f}")

          # negative dtw

          y = non_duplicates_z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][2])
          axes[1][2].set_title(f"negative DTW = {res.normalized_distance:.4f}")

```

Out [242]: Text(0.5,1,'negative DTW = 0.4641')



## 6 2.1 DTW distribution in latent space z (open\_end=True)

In [243]: # initialize dist\_lists

```
duplicates_dtw_z = []
non_duplicates_dtw_z = []

for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = z[i].squeeze()

    y_duplicate = duplicates_z[i].squeeze()
    duplicates_dtw_z.append(dtwalign.dtw(x, y_duplicate, dist=euclidean, open_end=True))

    y_non_duplicate = non_duplicates_z[i].squeeze()
    non_duplicates_dtw_z.append(dtwalign.dtw(x, y_non_duplicate, dist=euclidean, open_end=True))
```

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In [244]: bins = np.linspace(0, 1., 100)  
fig, axes = plt.subplots(1, 1, figsize=(15,8))

```

sns.distplot(duplicates_dtw_z, bins=bins, ax=axes, label="Duplicates")
axes.title.set_text('duplicates DTW')

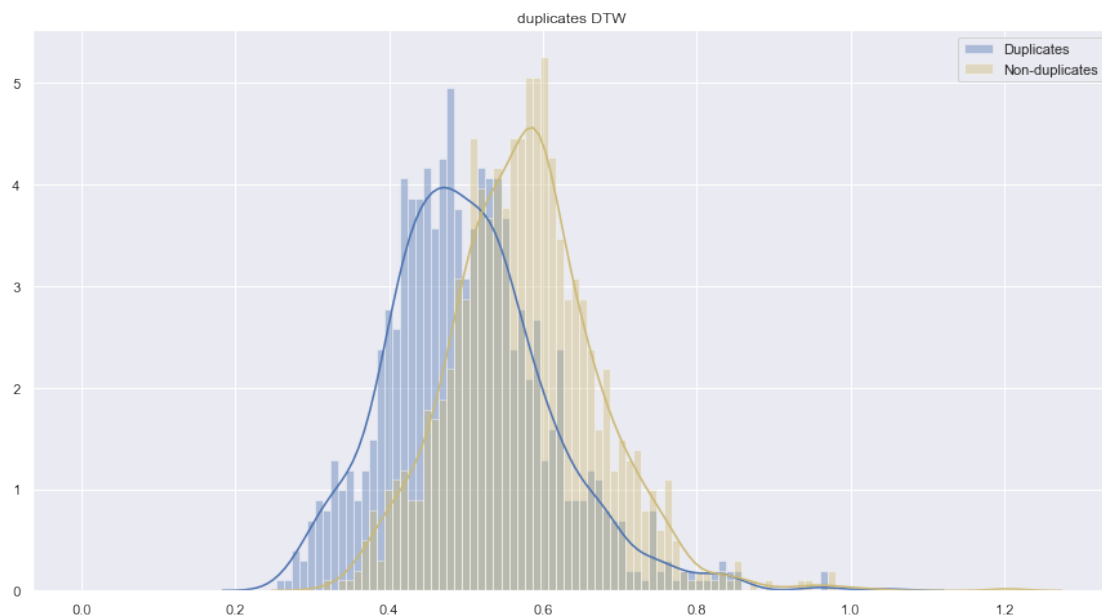
sns.distplot(non_duplicates_dtw_z, bins=bins, color='y', ax=axes, label='Non-duplicates')
axes.legend()
# axes[1].title.set_text('non-duplicates DTW')
plt.show()

hist_dup, _ = np.histogram(duplicates_dtw_z, bins=bins)
hist_dup = hist_dup / np.sum(hist_dup)

hist_non_dup, _ = np.histogram(non_duplicates_dtw_z, bins=bins)
hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
print(f"DKL is: {dkl:.4f}")

```



DKL is: 0.5891

## 7 2.2 DTW distribution in latent space z (open\_end=False)

```
In [245]: # initialize dist_lists
```

```
duplicates_dtw_z = []
```

```

non_duplicates_dtw_z = []

for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = z[i].squeeze()

    y_duplicate = duplicates_z[i].squeeze()
    duplicates_dtw_z.append(dtwalign.dtw(x, y_duplicate, dist=euclidean, open_end=False))

    y_non_duplicate = non_duplicates_z[i].squeeze()
    non_duplicates_dtw_z.append(dtwalign.dtw(x, y_non_duplicate, dist=euclidean, open_end=False))

```

999

```

In [246]: bins = np.linspace(0, 1.5, 100)
fig, axes = plt.subplots(1, 1, figsize=(15,8))

sns.distplot(duplicates_dtw_z, bins=bins, ax=axes, label="Duplicates")
axes.title.set_text('duplicates DTW')

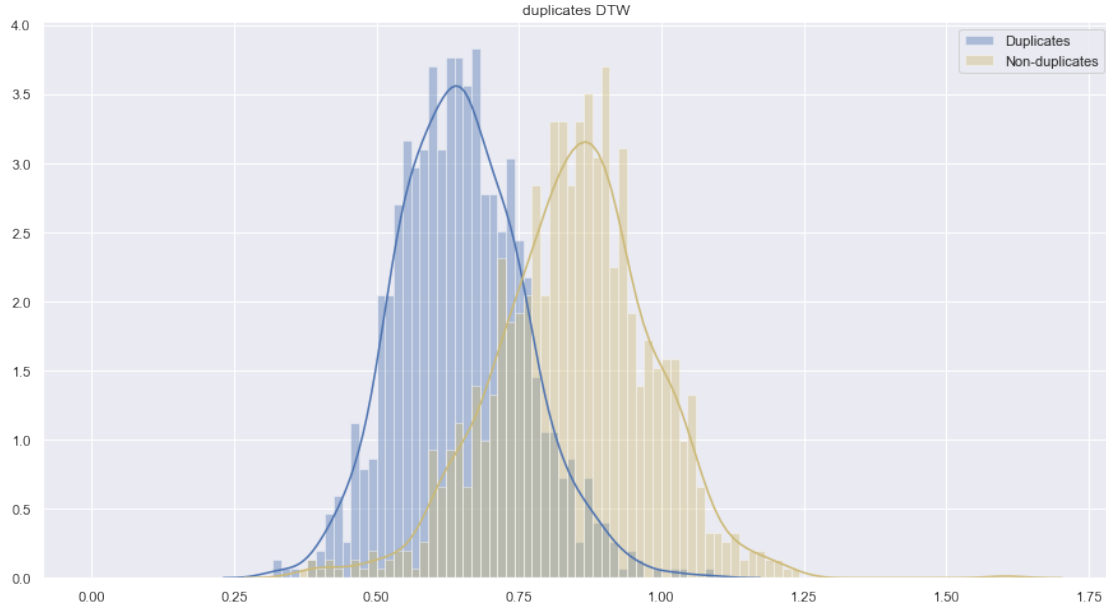
sns.distplot(non_duplicates_dtw_z, bins=bins, color='y', ax=axes, label='Non-duplicates DTW')
axes.legend()
# axes[1].title.set_text('non-duplicates DTW')
plt.show()

hist_dup, _ = np.histogram(duplicates_dtw_z, bins=bins)
hist_dup = hist_dup / np.sum(hist_dup)

hist_non_dup, _ = np.histogram(non_duplicates_dtw_z, bins=bins)
hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
print(f"DKL is: {dkl:.4f}")

```



DKL is: 1.2739

## 8 3. DTW (prefix case $n \ll m$ ) $n = 30$

```
In [247]: PREFIX_LEN = 30
```

```
In [248]: nn_input = torch.from_numpy(data['x'][:, :PREFIX_LEN, :]).cuda().float()
```

```
In [249]: z, *_ = nn.single_forward(nn_input)
          z = z.detach().cpu().numpy()
```

```
duplicates_z, *_ = nn.single_forward(torch.from_numpy(duplicates['x']).cuda().float())
duplicates_z = duplicates_z.detach().cpu().numpy()
```

```
non_duplicates_z, *_ = nn.single_forward(torch.from_numpy(non_duplicates['x']).cuda())
non_duplicates_z = non_duplicates_z.detach().cpu().numpy()
```

```
In [250]: # random latent variables
          # z, _, _, _ = nn.single_forward(nn_input)
          # z = z.detach().cpu().numpy()

          # duplicates_z, _, _, _ = nn.single_forward(torch.from_numpy(duplicates['x']).cuda())
          # duplicates_z = duplicates_z.detach().cpu().numpy()

          # non_duplicates_z, _, _, _ = nn.single_forward(torch.from_numpy(non_duplicates['x']).cuda())
          # non_duplicates_z = non_duplicates_z.detach().cpu().numpy()
```



```

In [251]: i = np.random.randint(data['y'].shape[0])
          sample_idx = i

          fig, axes = plt.subplots(2, 3, figsize=(15,10))

          # plt.figure()
          im = axes[0][0].imshow(z[i].T, vmin=-1, vmax=1)
          # fig.colorbar(im)
          label = index_to_word[data['y'][i]]
          axes[0][0].title.set_text(f'{label}')

          axes[0][1].imshow(duplicates_z[i].T, vmin=-1, vmax=1)
          # plt.colorbar()
          label = index_to_word[data['y'][i]]
          axes[0][1].title.set_text(f'{label}')

          axes[0][2].imshow(non_duplicates_z[i].T, vmin=-1, vmax=1)
          # plt.colorbar()
          label = index_to_word[non_duplicates['y'][i]]
          axes[0][2].title.set_text(f'{label}')

          ##### DTW

          x = z[sample_idx]

          # self dtw (expected to be 0)

          y = z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][0])
          axes[1][0].set_title(f"anchor DTW = {res.normalized_distance:.4f}")

          # positive dtw

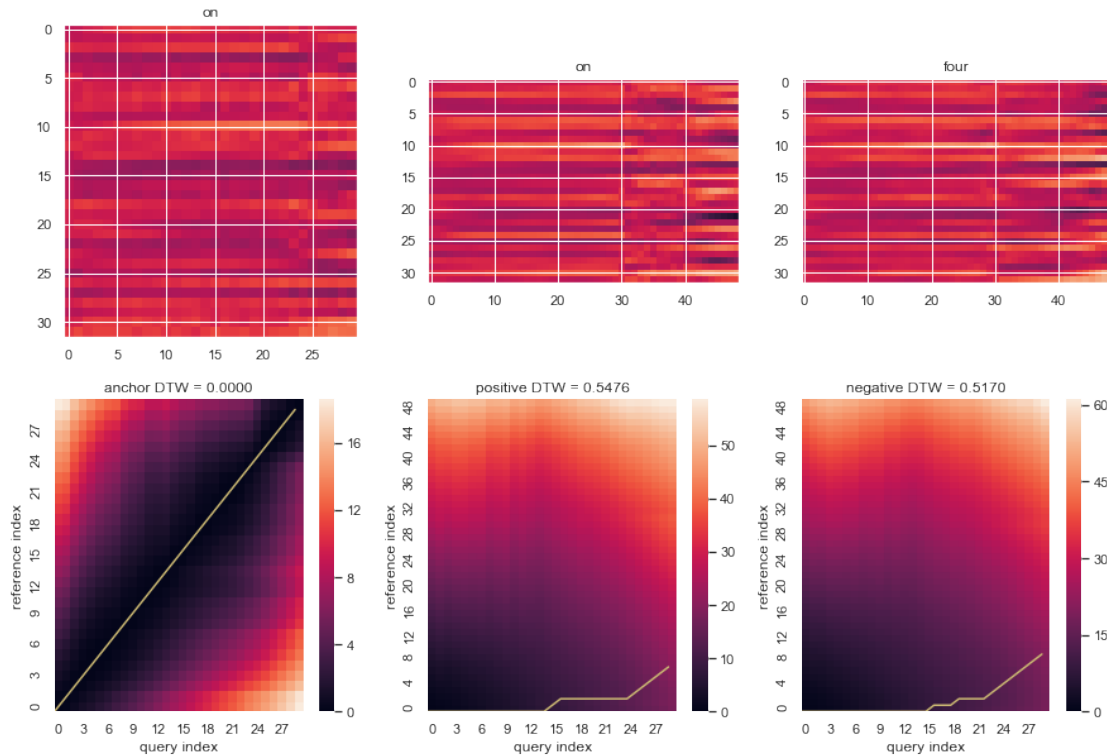
          y = duplicates_z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][1])
          axes[1][1].set_title(f"positive DTW = {res.normalized_distance:.4f}")

          # negative dtw

          y = non_duplicates_z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][2])
          axes[1][2].set_title(f"negative DTW = {res.normalized_distance:.4f}")

```

```
Out[251]: Text(0.5,1,'negative DTW = 0.5170')
```



## 9 3.1 DTW distribution in latent space z (open\_end=True)

```
In [252]: # initialize dist_lists
```

```
duplicates_dtw_z = []
non_duplicates_dtw_z = []

for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = z[i].squeeze()

    y_duplicate = duplicates_z[i].squeeze()
    duplicates_dtw_z.append(dtwalgn.dtw(x, y_duplicate, dist=euclidean, open_end=True))

    y_non_duplicate = non_duplicates_z[i].squeeze()
    non_duplicates_dtw_z.append(dtwalgn.dtw(x, y_non_duplicate, dist=euclidean, open_end=True))
```

999

```

In [253]: bins = np.linspace(0, 1., 100)
fig, axes = plt.subplots(1, 1, figsize=(15,8))

sns.distplot(duplicates_dtw_z, bins=bins, ax=axes, label="Duplicates")
axes.title.set_text('duplicates DTW')

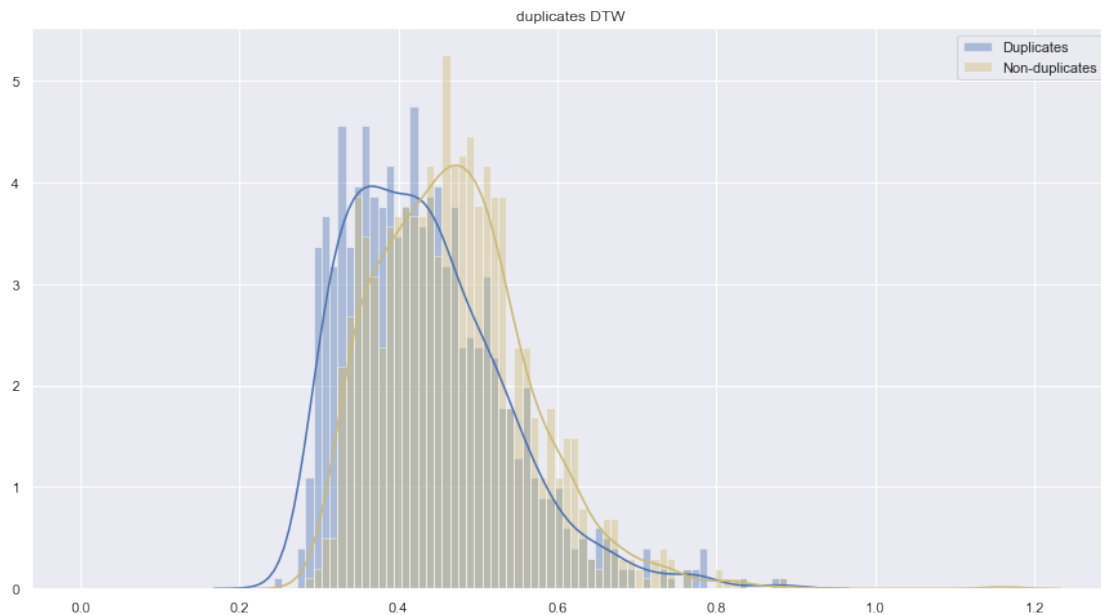
sns.distplot(non_duplicates_dtw_z, bins=bins, color='y', ax=axes, label='Non-duplicates')
axes.legend()
# axes[1].title.set_text('non-duplicates DTW')
plt.show()

hist_dup, _ = np.histogram(duplicates_dtw_z, bins=bins)
hist_dup = hist_dup / np.sum(hist_dup)

hist_non_dup, _ = np.histogram(non_duplicates_dtw_z, bins=bins)
hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
print(f"DKL is: {dkl:.4f}")

```



DKL is: 0.2957

## 10 3.2 DTW distribution in latent space z (open\_end=False)

```

In [254]: # initialize dist_lists

```

```

duplicates_dtw_z = []
non_duplicates_dtw_z = []

for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = z[i].squeeze()

    y_duplicate = duplicates_z[i].squeeze()
    duplicates_dtw_z.append(dtwalign.dtw(x, y_duplicate, dist=euclidean, open_end=False))

    y_non_duplicate = non_duplicates_z[i].squeeze()
    non_duplicates_dtw_z.append(dtwalign.dtw(x, y_non_duplicate, dist=euclidean, open_end=False))

```

999

```

In [255]: bins = np.linspace(0, 1.5, 100)
fig, axes = plt.subplots(1, 1, figsize=(15,8))

sns.distplot(duplicates_dtw_z, bins=bins, ax=axes, label="Duplicates")
axes.title.set_text('duplicates DTW')

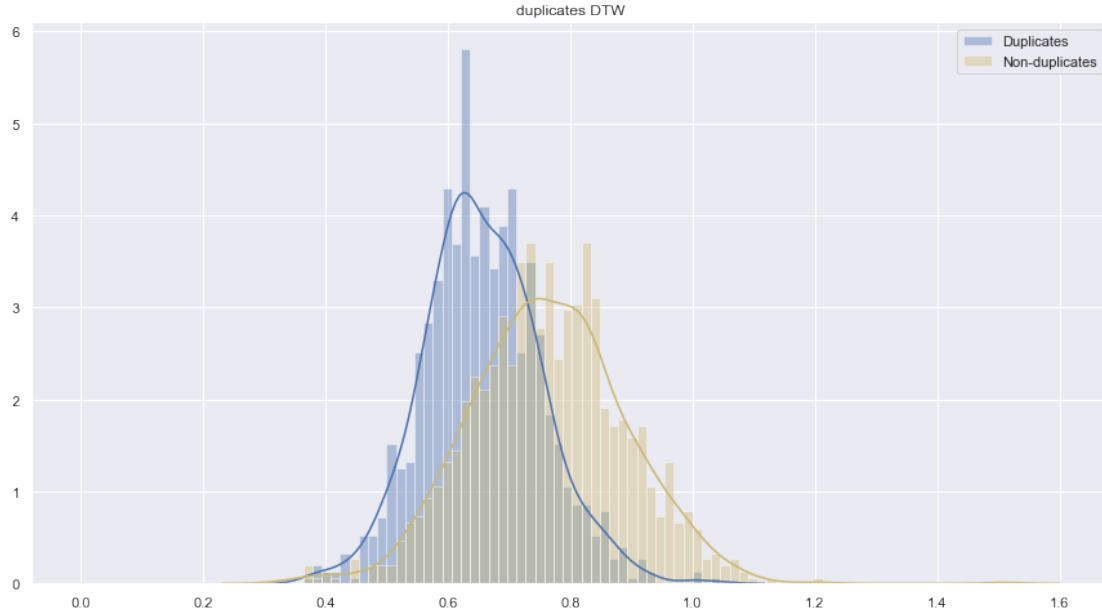
sns.distplot(non_duplicates_dtw_z, bins=bins, color='y', ax=axes, label='Non-duplicates DTW')
axes.legend()
# axes[1].title.set_text('non-duplicates DTW')
plt.show()

hist_dup, _ = np.histogram(duplicates_dtw_z, bins=bins)
hist_dup = hist_dup / np.sum(hist_dup)

hist_non_dup, _ = np.histogram(non_duplicates_dtw_z, bins=bins)
hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
print(f"DKL is: {dkl:.4f}")

```



DKL is: 0.5455

## 11 4. DTW (prefix case $n \ll m$ ) $n = 20$

```
In [256]: PREFIX_LEN = 20
```

```
In [257]: nn_input = torch.from_numpy(data['x'][:, :PREFIX_LEN, :]).cuda().float()
```

```
In [258]: z, *_ = nn.single_forward(nn_input)
          z = z.detach().cpu().numpy()
```

```
duplicates_z, *_ = nn.single_forward(torch.from_numpy(duplicates['x']).cuda().float())
duplicates_z = duplicates_z.detach().cpu().numpy()
```

```
non_duplicates_z, *_ = nn.single_forward(torch.from_numpy(non_duplicates['x']).cuda())
non_duplicates_z = non_duplicates_z.detach().cpu().numpy()
```

```
In [259]: # random latent variables
          # z, _, _, _ = nn.single_forward(nn_input)
          # z = z.detach().cpu().numpy()

          # duplicates_z, _, _, _ = nn.single_forward(torch.from_numpy(duplicates['x']).cuda())
          # duplicates_z = duplicates_z.detach().cpu().numpy()

          # non_duplicates_z, _, _, _ = nn.single_forward(torch.from_numpy(non_duplicates['x']).cuda())
          # non_duplicates_z = non_duplicates_z.detach().cpu().numpy()
```

```

In [260]: i = np.random.randint(data['y'].shape[0])
          sample_idx = i

          fig, axes = plt.subplots(2, 3, figsize=(15,10))

          # plt.figure()
          im = axes[0][0].imshow(z[i].T, vmin=-1, vmax=1)
          # fig.colorbar(im)
          label = index_to_word[data['y'][i]]
          axes[0][0].title.set_text(f'{label}')

          axes[0][1].imshow(duplicates_z[i].T, vmin=-1, vmax=1)
          # plt.colorbar()
          label = index_to_word[data['y'][i]]
          axes[0][1].title.set_text(f'{label}')

          axes[0][2].imshow(non_duplicates_z[i].T, vmin=-1, vmax=1)
          # plt.colorbar()
          label = index_to_word[non_duplicates['y'][i]]
          axes[0][2].title.set_text(f'{label}')

          ##### DTW

          x = z[sample_idx]

          # self dtw (expected to be 0)

          y = z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][0])
          axes[1][0].set_title(f"anchor DTW = {res.normalized_distance:.4f}")

          # positive dtw

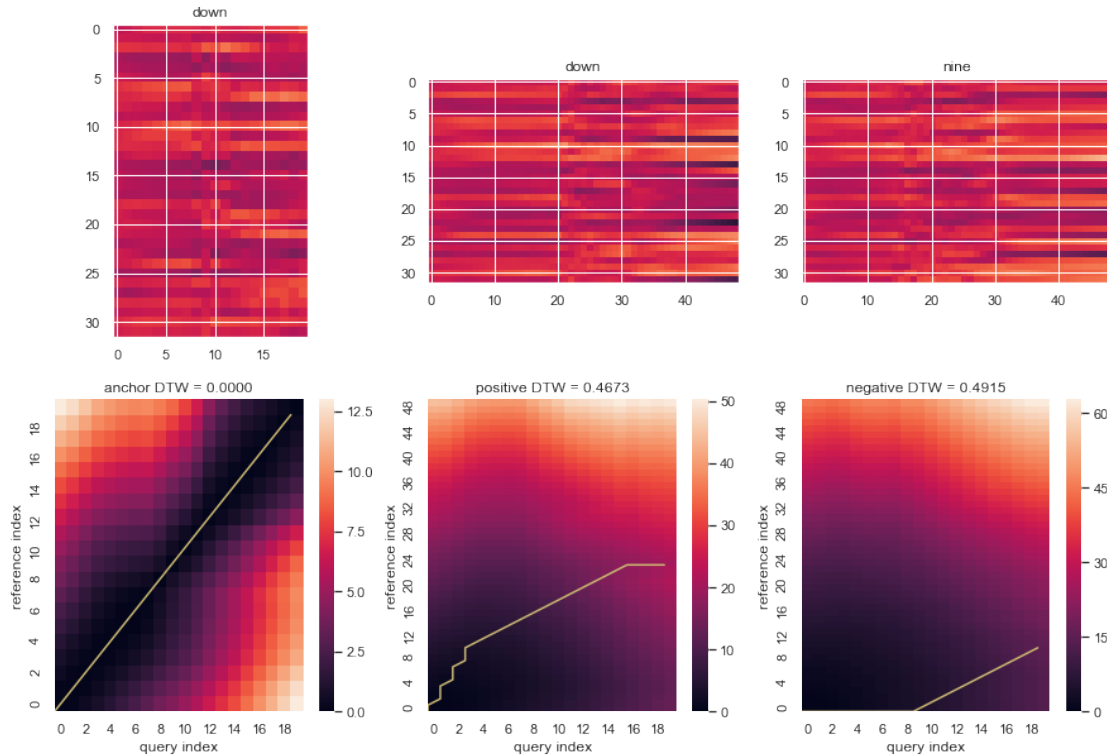
          y = duplicates_z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][1])
          axes[1][1].set_title(f"positive DTW = {res.normalized_distance:.4f}")

          # negative dtw

          y = non_duplicates_z[sample_idx]
          res = dtwalign.dtw(x, y, dist=euclidean, open_end=True)
          res.plot_path(ax=axes[1][2])
          axes[1][2].set_title(f"negative DTW = {res.normalized_distance:.4f}")

```

Out[260]: Text(0.5,1,'negative DTW = 0.4915')



## 12 4.1 DTW distribution in latent space z (open\_end=True)

In [261]: # initialize dist\_lists

```
duplicates_dtw_z = []
non_duplicates_dtw_z = []

for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = z[i].squeeze()

    y_duplicate = duplicates_z[i].squeeze()
    duplicates_dtw_z.append(dtwalgn.dtw(x, y_duplicate, dist=euclidean, open_end=True))

    y_non_duplicate = non_duplicates_z[i].squeeze()
    non_duplicates_dtw_z.append(dtwalgn.dtw(x, y_non_duplicate, dist=euclidean, open_end=True))
```

999

```

In [262]: bins = np.linspace(0, 1., 100)
fig, axes = plt.subplots(1, 1, figsize=(15,8))

sns.distplot(duplicates_dtw_z, bins=bins, ax=axes, label="Duplicates")
axes.title.set_text('duplicates DTW')

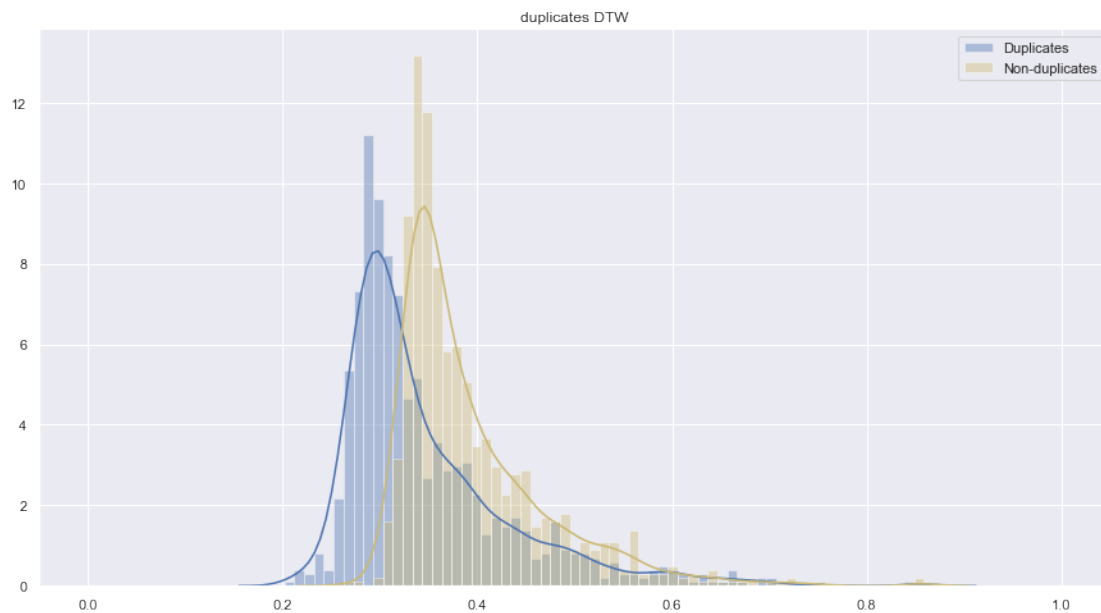
sns.distplot(non_duplicates_dtw_z, bins=bins, color='y', ax=axes, label='Non-duplicates')
axes.legend()
# axes[1].title.set_text('non-duplicates DTW')
plt.show()

hist_dup, _ = np.histogram(duplicates_dtw_z, bins=bins)
hist_dup = hist_dup / np.sum(hist_dup)

hist_non_dup, _ = np.histogram(non_duplicates_dtw_z, bins=bins)
hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
print(f"DKL is: {dkl:.4f}")

```



DKL is: 2.1786

## 13 4.2 DTW distribution in latent space z (open\_end=False)

```

In [263]: # initialize dist_lists

```



```

duplicates_dtw_z = []
non_duplicates_dtw_z = []

for i in range(n_mini_batch_size):
    print(i, end='\r')
    x = z[i].squeeze()

    y_duplicate = duplicates_z[i].squeeze()
    duplicates_dtw_z.append(dtwalign.dtw(x, y_duplicate, dist=euclidean, open_end=False))

    y_non_duplicate = non_duplicates_z[i].squeeze()
    non_duplicates_dtw_z.append(dtwalign.dtw(x, y_non_duplicate, dist=euclidean, open_end=False))

```

999

```

In [264]: bins = np.linspace(0, 1.5, 100)
fig, axes = plt.subplots(1, 1, figsize=(15,8))

sns.distplot(duplicates_dtw_z, bins=bins, ax=axes, label="Duplicates")
axes.title.set_text('duplicates DTW')

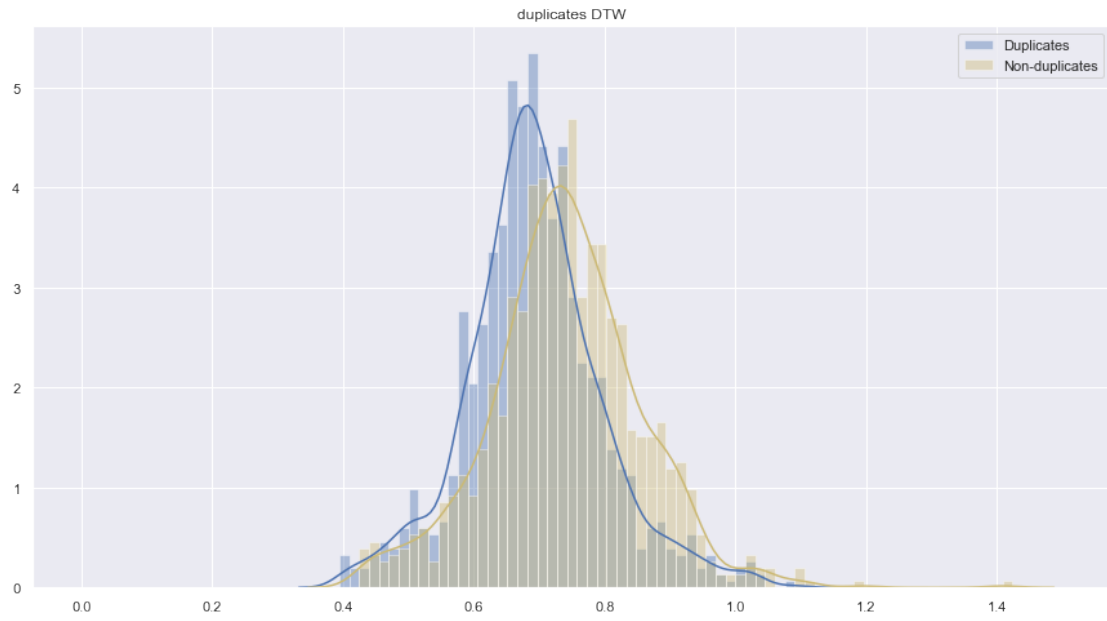
sns.distplot(non_duplicates_dtw_z, bins=bins, color='y', ax=axes, label='Non-duplicates DTW')
axes.legend()
# axes[1].title.set_text('non-duplicates DTW')
plt.show()

hist_dup, _ = np.histogram(duplicates_dtw_z, bins=bins)
hist_dup = hist_dup / np.sum(hist_dup)

hist_non_dup, _ = np.histogram(non_duplicates_dtw_z, bins=bins)
hist_non_dup = hist_non_dup / np.sum(hist_non_dup)

dkl = scipy.special.kl_div(hist_dup+1e-5, hist_non_dup+1e-5).sum()
print(f"DKL is: {dkl:.4f}")

```



DKL is: 0.1969