eval rna bubble

September 2, 2024

```
[]: import numpy as np
     import os
     import json
     import re
     import joblib
     import matplotlib.pyplot as plt
     import matplotlib.colors as plt_colors
     import tensorflow as tf
     import tensorflow.keras as keras
     import smfret.fit as fit
     from smfret.tf_layers import Attention
     from smfret.tf_layers import Conv
     from smfret.tf_layers import Summary
     from smfret.tf_layers import PrependTaskToken
     from smfret.tf_layers import Embedding
     from smfret.tf_layers import PositionEmbedding
     from smfret.trace_simulator import Simulator
     from smfret.trace simulator import ParameterGenerator
     from smfret.trace_simulator import SimulatedTraceSet
     from smfret.multi_task_learning import FRETStateTraceSet
     from smfret.multi_task_learning import TwoStateQuickDynamicTraceSet
     from smfret.dataset import MatlabTraceSet
     from smfret.dataset import FRETTraceSet
     from sklearn.manifold import TSNE
     from sklearn import metrics
     import sklearn
     from sklearn import mixture
     import functools
     import umap
     import scipy
```

```
from sciplotlib import style as spstyle
    from matplotlib.colors import LogNorm
    import matplotlib.cm as cm
[2]: rng = np.random.default_rng()
[3]: keras.mixed precision.set global policy('mixed bfloat16')
     encoder = keras.models.load_model('../../saved_models/best_model/
      ⇔encoder-20240111-045226.h5', compile=False)
[4]: def interpolate_position_embedding(position_embedding, periodicity=40):
         """Interpolates the position embedding vectors."""
        vectors = []
        w = position_embedding.variables[0]
        for i in range(w.shape[0]):
            if i == 0:
                vectors.append(w[i, :])
            elif i < periodicity:</pre>
                vectors.append(w[i, :])
            else:
                vectors.append(w[i - periodicity + 1, :])
        position_embedding.variables[0].assign(tf.stack(vectors, axis=0))
[5]: position_embedding = encoder.layers[-1].transformer.position_embedding
     interpolate position embedding(position embedding, periodicity=40)
[6]: NN_K = 50 # how many nearest neighbors are considered for the score
[7]: def calculate_distance(embedding):
         """Calculates the distance matrix from the embeddings."""
        distance = (
            tf.expand_dims(tf.einsum('ik,ik->i', embedding, embedding), axis=-1)
            + tf.expand_dims(tf.einsum('ik,ik->i', embedding, embedding), axis=0)
            - 2.0 * tf.einsum('ik,jk->ij', embedding, embedding))
        return distance
[8]: def get_perplexity(embedding, label, branching_factor,_
      """Calculates the perplexity of a trace's neighbors."""
        distance = calculate_distance(embedding)
        top_k_results = tf.math.top_k(
            -distance, k=branching_factor + 1, sorted=True, name=None
        top_k_idx = top_k_results.indices.numpy()
        n = embedding.shape[0]
        perplexity = []
        for i in range(n):
```

```
this_label = label[i]
neighbour_label = label[top_k_idx[i, :].flatten()]
unique_labels = set(neighbour_label)
shannon_entropy = 0
for l in unique_labels:
    p = np.mean(neighbour_label == 1)
    shannon_entropy -= p * np.log(p)
    perplexity.append(shannon_entropy)
return np.array(perplexity)
```

```
[12]: def generate embedding and label(trace sets, start frame, encoder,
       →max_frame=2000, max_traces=4000, balance_labels=True):
          """Generates the embeddings and labels."""
          embeddings = []
          labels = []
          colors = []
          color count = 1
          image_files = []
          traces = []
          count = 0
          for trace_set in trace_sets:
              trim_size = len(trace_set.time) // 100 * 100
              trace_set.trim(trim_size)
              trace_set.trim(max_frame, start_frame=start_frame)
              trace_set.broadcast_data_to_traces()
              print(trace_set.size)
              if count + trace_set.size > max_traces:
                  break
              else:
                  count += trace_set.size
              with tf.device('/CPU:0'):
                  label = np.max(trace_set.label, axis=-1)
                  if balance_labels:
                      n_included = np.sum(label == 1)
                      n_excluded = np.sum(label == 0)
                      if n_included == 0 or n_excluded == 0:
                          continue
                      indices = np.arange(len(label))
                      ratio = int(n_excluded / n_included)
                      # ratio = 0
                      if ratio > 1:
                          indices = np.concatenate([indices[label == 1],__
       →indices[label ==0][::ratio]])
                          label = label[indices]
```

```
trace_set = trace_set.get_subset(indices)
              embedding = encoder.predict(trace_set.to_tensor(), batch_size=8)
              embeddings.append(embedding)
              labels.append(label)
              traces.extend(trace_set.traces)
       return embeddings, labels
[13]: config_file = '../../experiment_conditions/cond_rna_bubble_all.json'
    with open(config_file) as f:
        config = json.load(f)
    trace_sets = [MatlabTraceSet('.../.../' + file) for file in_
     ⇔config['expt_setting_set_files'].strip().split('\n')]
    embeddings, labels = generate_embedding_and_label(trace_sets,_
     -config['start_frame'], encoder, max_frame=2000, max_traces=20000, __
     ⇔balance labels=False)
    504
    63/63 [=========== ] - 4s 53ms/step
    65/65 [========] - 3s 54ms/step
    537
    68/68 [========] - 4s 55ms/step
    61/61 [======== ] - 3s 53ms/step
    550
    69/69 [=======] - 4s 59ms/step
    468
    59/59 [=======] - 3s 57ms/step
    58/58 [========= ] - 3s 56ms/step
    55/55 [========= ] - 3s 58ms/step
    72/72 [========= ] - 4s 54ms/step
    519
    65/65 [======== ] - 4s 54ms/step
    68/68 [========= ] - 4s 53ms/step
    65/65 [========= ] - 3s 52ms/step
    63/63 [========] - 3s 53ms/step
    543
    68/68 [======== ] - 4s 54ms/step
    56/56 [========= ] - 3s 54ms/step
```

454

```
58/58 [======== ] - 3s 53ms/step
    53/53 [========= ] - 3s 53ms/step
    53/53 [========= ] - 3s 53ms/step
    52/52 [========] - 3s 53ms/step
    417
    53/53 [========= ] - 3s 52ms/step
    49/49 [======== ] - 3s 53ms/step
    50/50 [======== ] - 3s 54ms/step
    47/47 [========] - 3s 59ms/step
    49/49 [======== ] - 3s 53ms/step
    381
    48/48 [========= ] - 3s 54ms/step
    340
    43/43 [========= ] - 1s 17ms/step
    46/46 [========] - 3s 60ms/step
    44/44 [========] - 2s 54ms/step
    424
    53/53 [========= ] - 3s 54ms/step
[14]: final_embeddings = []
    final_labels = []
    human_labels = []
    for e, t, l in zip(embeddings, trace_sets, labels):
       final_labels.append([t.file.split('/')[-2]] * e.shape[0])
       final_embeddings.append(e)
       human_labels.append(1)
    embedding = np.concatenate(final_embeddings, axis=0)
    label = np.concatenate(final_labels)
    human_label = np.concatenate(human_labels)
[17]: ordered_label_spot = sorted(list(set(label.tolist())), key=lambda s: int(re.

→findall(string=s, pattern='\d+')[0]))
[18]: perplexity_spot = perplexity
    new_label_spot = label
    embedding_spot = embedding
```

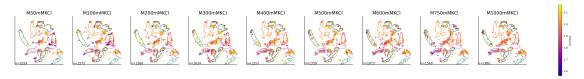
57/57 [========] - 3s 54ms/step

```
[19]: cutoff = np.quantile(perplexity_spot, 1.0)
[21]: reducer = joblib.load('../../saved_models/UMAP/atlas_pp.joblib')
      saved_atlas_density_models = joblib.load('../../saved_models/UMAP/
       ⇔atlas_pp_density_models.joblib')
      cutoff = np.quantile(perplexity_spot, 1.0)
      counter = 0
      z_h = reducer.transform(embedding_spot)
      ax min = np.min(z h)
      ax_max = np.max(z_h)
      with plt.style.context(spstyle.get_style('nature-reviews')):
          fig, axes = plt.subplots(ncols=len(ordered_label_spot), nrows=1, figsize=(5_U

+* len(ordered_label_spot), 5));
      for 1 in ordered label spot:
          idx = (new label spot == 1) & (perplexity spot <= cutoff)</pre>
          z = z h[idx, ...]
          with plt.style.context(spstyle.get_style('nature-reviews')):
              ax = axes[counter]
              counter += 1
              s = ax.scatter(z[:, 0], z[:, 1], c=perplexity_spot[idx], s=30, alpha=0.
       →6, linewidths=0, cmap='plasma', vmin=np.quantile(perplexity_spot, 0.01), u
       →vmax=np.quantile(perplexity_spot, 0.99))
              ax.set xlim([ax min, ax max])
              ax.set_ylim([ax_min, ax_max])
              ax.set_xticks([])
              ax.set_yticks([])
              ax.set_aspect('equal')
              ax.set_title(f'{1}')
              ax.text(x=ax.get_xlim()[0] + 0.1, y=ax.get_ylim()[0] + 0.1, s=f'n={z.}
       ⇒shape[0]}', fontdict={'size': 12})
          contour counter = 0
          for 1 in saved_atlas_density_models:
              color = plt.cm.tab20.colors + plt.cm.tab20b.colors[::2]
              x = np.linspace(np.min(z_h[:, 0]) - 1, np.max(z_h[:, 0]) + 1, num=200)
              y = np.linspace(np.min(z_h[:, 1]) - 1, np.max(z_h[:, 1]) + 1, num=200)
              X, Y = np.meshgrid(x, y)
              clf = saved_atlas_density_models[1]
              XX = np.array([X.ravel(), Y.ravel()]).T
              Z = clf.score_samples(XX)
              Z = np.exp(Z.reshape(X.shape))
              cmap = plt_colors.LinearSegmentedColormap.from_list("",__
       →[color[contour_counter], color[contour_counter], color[contour_counter]])
              if 'n' not in 1:
                  CS = ax.contour(
                      X, Y, Z, cmap=cmap, levels=[1e-2]
```

```
contour_counter += 1
color_bar = fig.colorbar(s, ax=axes, label="Entropy")
color_bar.set_alpha(1)
color_bar.draw_all()
if cutoff >= np.quantile(perplexity_spot, 0.9):
    plt.savefig('figures/rna_bubble_perplexity_umap_atlas.svg')
else:
    plt.savefig('figures/rna_bubble_perplexity_umap_atlas_threshold.svg')
plt.show()
```

OMP: Info #276: omp_set_nested routine deprecated, please use omp_set_max_active_levels instead.



```
[22]: reducer = umap.UMAP(n_epochs=500, n_neighbors=30, min_dist=0.0, random_state=np.
       →random.RandomState(60), n_components=2, negative_sample_rate=15)
      reducer.fit transform(embedding spot)
      counter = 0
      z h = reducer.transform(embedding spot)
      ax_min = np.min(z_h)
      ax_max = np.max(z_h)
      with plt.style.context(spstyle.get_style('nature-reviews')):
          fig, axes = plt.subplots(ncols=len(ordered_label_spot), nrows=1, figsize=(5_

state = * len(ordered_label_spot), 5));

      for l in ordered label spot:
          idx = (new_label_spot == 1) & (perplexity_spot <= cutoff)</pre>
          z = z_h[idx, ...]
          with plt.style.context(spstyle.get_style('nature-reviews')):
              ax = axes[counter]
              counter += 1
              if cutoff >= np.quantile(perplexity_spot, 0.9):
                  s = ax.scatter(z[:, 0], z[:, 1], c=perplexity_spot[idx], s=8,__
       ⇒alpha=1, linewidths=0, cmap='plasma', vmin=np.quantile(perplexity_spot, 0.
       ⇔01), vmax=np.quantile(perplexity_spot, 0.99))
              else:
                  s = ax.scatter(z[:, 0], z[:, 1], c=perplexity_spot[idx], s=30, __
       →alpha=0.5, linewidths=0, cmap='plasma', vmin=np.quantile(perplexity_spot, 0.
       →01), vmax=np.quantile(perplexity_spot, 0.99))
```

```
ax.set_xlim([ax_min, ax_max])
        ax.set_ylim([ax_min, ax_max])
        ax.set_xticks([])
        ax.set_yticks([])
        ax.set_aspect('equal')
        ax.set_title(f'{1}')
        ax.text(x=ax.get_xlim()[0] + 0.1, y=ax.get_ylim()[0] + 0.1, s=f'n={z.}
 ⇔shape[0]}', fontdict={'size': 12})
color_bar = fig.colorbar(s, ax=axes, label="Entropy")
color_bar.set_alpha(1)
color_bar.draw_all()
if cutoff >= np.quantile(perplexity_spot, 0.9):
    plt.savefig('figures/rna_bubble_perplexity_umap_contour.svg')
else:
    plt.savefig('figures/rna_bubble_perplexity_umap_threshold.svg')
plt.show()
```

