

# CS5340 - Lab3 Report

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## 1. E-step

**Emission probability** we first compute emission probability with *scipy.stats.norm.pdf* function.

$$emission[i, n, k] = p(x_n^{(i)} | z_{nk} = 1, \mu_k, \sigma_k)$$

Here  $i = 1, 2, \dots, T$  is the index of observed sequence,  $n = 1, 2, \dots, N$  is the index of node inside each sequence, and  $k = 1, 2, \dots, K$  refers to the hidden state.

$\hat{\alpha}^{(i)}(z_1)$  and  $\hat{\beta}^{(i)}(z_N)$  initialize  $\hat{\alpha}^{(i)}(z_1) = \frac{\alpha^{(i)}(z_1)}{c_1}$ , with  $\alpha^{(i)}(z_1) = p(x_1^{(i)}, z_1) = p(x_1^{(i)} | z_1)p(z_1)$  and normalizer  $c_1 = \sum_{z_1} \alpha^{(i)}(z_1)$ , and initialize  $\hat{\beta}^{(i)}(z_N) = \mathbf{1}$

**forward process** for  $n = 2$  to  $N$ , compute  $\hat{\alpha}^{(i)}(z_n), c_n$  with

$$c_n \hat{\alpha}^{(i)}(z_n) = p(x_n^{(i)} | z_n) \sum_{z_{n-1}} \hat{\alpha}^{(i)}(z_{n-1}) p(z_n | z_{n-1}) := \tilde{\alpha}^{(i)}(z_n)$$
$$c_n = \sum_{z_n} \tilde{\alpha}^{(i)}(z_n)$$

**backward process** for  $n = N-1$  to  $1$ , compute  $\hat{\beta}^{(i)}(z_n)$  with

$$c_{n+1} \hat{\beta}^{(i)}(z_n) = \sum_{z_{n+1}} p(x_{n+1}^{(i)} | z_{n+1}) p(z_{n+1} | z_n) \hat{\beta}^{(i)}(z_{n+1}) := \tilde{\beta}^{(i)}(z_n)$$

**compute  $\gamma$  and  $\xi$**  with

$$\gamma^{(i)}(z_n) = \hat{\alpha}^{(i)}(z_n) \hat{\beta}^{(i)}(z_n)$$
$$\xi^{(i)}(z_{n-1}, z_n) = \frac{\hat{\alpha}^{(i)}(z_{n-1}) p(x_n^{(i)} | z_n) p(z_n | z_{n-1}) \hat{\beta}^{(i)}(z_n)}{c_n}$$

## 2. M-step

**update  $\pi$**  with

$$\pi_k = \frac{\sum_{i=1}^T \gamma^{(i)}(z_{1k})}{T}$$

**update  $A$**  with

$$A_{jk} = \frac{\sum_{i=1}^T \sum_{n=2}^N \xi^{(i)}(z_{n-1,j}, z_{nk})}{\sum_{i=1}^T \sum_{n=2}^N \sum_{l=1}^K \xi^{(i)}(z_{n-1,j}, z_{nl})}$$

**update  $\mu_k$  and  $\sigma_k$**  here  $x_n^{(i)} \in \mathbb{R}$  is sampled from Gaussian distribution  $\mathcal{N}(\mu_k, \sigma_k)$

$$\mu_k = \frac{\sum_{i=1}^T \sum_{n=1}^N \gamma^{(i)}(z_{nk}) x_n^{(i)}}{\sum_{i=1}^T \sum_{n=1}^N \gamma^{(i)}(z_{nk})}$$
$$\sigma_k = \frac{\sum_{i=1}^T \sum_{n=1}^N \gamma^{(i)}(z_{nk}) (x_n^{(i)} - \mu_k)^2}{\sum_{i=1}^T \sum_{n=1}^N \gamma^{(i)}(z_{nk})}$$

### 3. EM algorithm

**initialize**  $\pi, A, \mu, \sigma$  with K-means clustering

**update params** run E-step and M-step iteratively until convergence.

### 4. Results

```
yichao@macbook:~/Documents/CS5340/lab3$ python test_lab3.py
```

```
e_step: PASSED
```

```
m_step: PASSED
```

```
Running on seq_short
```

```
-----
```

```
Loaded 200 sequences, with average length = 8.0
```

```
Groundtruth pi:
```

```
  [0.9 0.1 0. ]
```

```
Groundtruth A:
```

```
  [[0.3 0.7 0. ]
```

```
   [0.  0.6 0.4]
```

```
   [0.  0.  1. ]]
```

```
Groundtruth phi:
```

```
  {'mu': array([-1.5,  0.5, -0.2]), 'sigma': array([0.5, 0.2, 0.3])}
```

```
Your pi:
```

```
  [0.  0.91 0.09]
```

```
Your A:
```

```
  [[1.  0.  0. ]
```

```
   [0.  0.28 0.72]
```

```
   [0.4 0.  0.6 ]]
```

```
Your phi:
```

```
  {'mu': array([-0.21, -1.5 ,  0.52]), 'sigma': array([0.29, 0.51, 0.21])}
```

```
Time running fit_hmm() on dataset seq_short : 1.21s
```

```
Running on seq_long
```

```
-----
```

```
Loaded 5 sequences, with average length = 1000.0
```

```
Groundtruth pi:
```

```
  [0.6 0.  0.  0.4]
```

```
Groundtruth A:
```

```
  [[0.3 0.  0.  0.7]
```

```
   [0.7 0.3 0.  0. ]
```

```
   [0.  0.6 0.4 0. ]
```

```
   [0.  0.  0.4 0.6]]
```

```
Groundtruth phi:
```

```
  {'mu': array([0. , 1.1, 2. , 1. ]), 'sigma': array([0.5, 0.3, 0.4, 0.2])}
```

```
Your pi:
```

```
  [0.64 0.  0.36 0. ]
```

```
Your A:
```

```
  [[0.3 0.  0.7 0. ]
```

```
   [0.7 0.3 0.  0. ]
```

```
   [0.  0.  0.57 0.43]
```

```
   [0.  0.58 0.  0.42]]
```

```
Your phi:
```

```
  {'mu': array([0. , 1.07, 1. , 2. ]), 'sigma': array([0.52, 0.31, 0.19, 0.41])}
```

```
Time running fit_hmm() on dataset seq_long : 3.56s
```

```
Running on seq_short2
```

```
-----
```

```

Loaded 100 sequences, with average length = 9.0
Groundtruth pi:
[0.4 0.3 0. 0.3]
Groundtruth A:
[[0.4 0.6 0. 0. ]
 [0.2 0.2 0.6 0. ]
 [0. 0. 0.1 0.9]
 [0. 0. 0. 1. ]]
Groundtruth phi:
{'mu': array([ 0. , -1. , 0.5, -0.5]), 'sigma': array([0.2, 0.2, 0.1, 0.1])}
Your pi:
[0.33 0.25 0.42 0. ]
Your A:
[[0.18 0. 0.2 0.62]
 [0. 1. 0. 0. ]
 [0.58 0. 0.42 0. ]
 [0. 0.96 0. 0.04]]
Your phi:
{'mu': array([-0.99, -0.5 , 0.01, 0.5 ]), 'sigma': array([0.22, 0.1 , 0.17, 0.09])}
Time running fit_hmm() on dataset seq_short2 : 0.45s

```

Running on seq\_long2

```

-----
Loaded 5 sequences, with average length = 800.0
Groundtruth pi:
[1. 0. 0. 0. 0. 0.]
Groundtruth A:
[[0.3 0.7 0. 0. 0. 0. ]
 [0. 0.6 0.4 0. 0. 0. ]
 [0.2 0. 0.3 0.5 0. 0. ]
 [0. 0. 0. 0.4 0.5 0.1]
 [0. 0. 0. 0. 0.8 0.2]
 [0.5 0. 0. 0.1 0. 0.4]]
Groundtruth phi:
{'mu': array([ 0.5, 0. , 2. , -1. , -0.2, 1.1]), 'sigma': array([0.5 , 0.25,
0.2 , 0.2 , 0.4 , 0.1 ])}
Your pi:
[0. 0. 0. 0. 1. 0.]
Your A:
[[0.6 0.4 0. 0. 0. 0. ]
 [0. 0.26 0.53 0. 0.2 0. ]
 [0. 0. 0.43 0.09 0. 0.47]
 [0. 0. 0.08 0.42 0.49 0. ]
 [0.69 0. 0. 0. 0.31 0. ]
 [0. 0. 0. 0.21 0. 0.79]]
Your phi:
{'mu': array([-0.01, 1.99, -0.98, 1.1 , 0.49, -0.21]), 'sigma': array([0.25, 0.19,
0.2 , 0.1 , 0.49, 0.39])}
Time running fit_hmm() on dataset seq_long2 : 3.33s

```