# EM algorithm for coin flipping problem 

## Cheng Li

We have $K$ coins. The mixing proportions are $\pi_{1}, \pi_{2}, \ldots, \pi_{K}$. The probability of the $k$ 's coin getting heads is $q_{k}$. First we randomly pick a coin, then we flip this coin for $d$ times. And we repeat this process for $N$ times. In this way, we can generate $N$ data points $\left\{x_{1}, x_{2}, \ldots, x_{N}\right\}$, each of which is a $d$ dimensional vector. $x_{i} \in\{\text { Head, Tail }\}^{d}$. Suppose the number of heads in $x_{k}$ is $y_{k}$. We define the hidden variables $z_{n}$, representing the component assignment for data point $x_{n}$ using a vector of size $K$. If $x_{n}$ is drawn from the $k$ th component, $z_{n k}=1$ while the remaining are all 0 .

- E step: Compute $\gamma\left(z_{n k}\right)$ with current parameters $\theta=\left\{\pi_{k}, q_{k}\right\}$.

$$
\gamma\left(z_{n k}\right)=p\left(z_{n k}=1 \mid x_{n}, \theta\right)=\frac{\pi_{k} p\left(x_{n} \mid q_{k}\right)}{\sum_{j=1}^{K} \pi_{j} p\left(x_{n} \mid q_{j}\right)}
$$

- M step: update $\pi_{k}$ and $q_{k}$

$$
\begin{gathered}
\pi_{k}=\frac{\sum_{n=1}^{N} \gamma\left(z_{n k}\right)}{N} \\
q_{k}=\frac{\sum_{n=1}^{N} \gamma\left(z_{n k}\right) y_{n}}{\sum_{n=1}^{N} \gamma\left(z_{n k}\right) d}
\end{gathered}
$$

## References

[1] Dawen Liang, Technical Details about the Expectation Maximization (EM) Algorithm, 2012

