

# Problem 1

$$\ln p(x|\pi, \theta) = \sum_c q(c) \ln \frac{p(x, c|\pi, \theta)}{q(c)} + \sum_c q(c) \ln \frac{q(c)}{p(c|\pi, \theta)}$$

E-step: Set  $q(c) = p(c|x, \pi, \theta)$ .

$$\begin{aligned} p(c|x, \pi, \theta) &\propto \prod_{i=1}^n p(x_i|G_i, \theta) p(G_i|\pi) \\ &= \prod_{i=1}^n \frac{p(x_i|G_i, \theta) p(G_i|\pi)}{\sum_j p(x_i|G_i=j, \theta) p(G_i=j|\pi)} = \prod_{i=1}^n p(G_i|x_i, \pi, \theta) = \prod_{i=1}^n q(G_i) \end{aligned}$$

where  $p(G_i=k|x_i, \pi, \theta) = \frac{p(x_i|G_i=k, \pi, \theta) p(G_i=k|\pi)}{\sum_{j=1}^K p(x_i|G_i=j, \pi, \theta) p(G_i=j|\pi)}$

$$= \frac{\pi_k \text{Binomial}(x_i|\theta_k)}{\sum_{j=1}^K \pi_j \text{Binomial}(x_i|\theta_j)}$$

$$\begin{aligned} \mathcal{L}(\pi, \theta) &= \sum_{i=1}^n \mathbb{E}_{q(c)} [\ln p(x_i, G_i|\pi, \theta)] + \text{const.} \\ &= \sum_{i=1}^n \sum_{j=1}^K \phi_i(j) [\ln p(x_i|G_i=j, \pi, \theta) + \ln p(G_i=j|\pi)] + \text{const.} \\ &= \sum_{i=1}^n \sum_{j=1}^K \phi_i(j) [\pi_j \ln \theta_j + (20 - x_i) \ln(1 - \theta_j) + \ln \pi_j] + \text{const.} \end{aligned}$$

M-step:  $\nabla_{\theta_j} \mathcal{L} = 0$

$$\theta_j = \frac{\sum_{i=1}^n \phi_i(j) x_i}{\sum_{i=1}^n \phi_i(j)} = \frac{\sum_{i=1}^n \phi_i(j) x_i}{20 n_j}, \quad n_j = \sum_{i=1}^n \phi_i(j)$$

$$\nabla_{\pi} \mathcal{L} = 0 \quad \pi_j = \frac{n_j}{n}, \quad n_j = \sum_{i=1}^n \phi_i(j)$$

$$\begin{aligned} f_t = \ln p(x|\pi, \theta) &= \cancel{\sum_{i=1}^n \sum_{j=1}^K \phi_i(j) \ln \pi_j + x_i \ln \theta_j} \\ &= \sum_{i=1}^n \sum_{j=1}^K \phi_i(j) \ln \pi_j + \ln \text{Binomial}(x_i|\theta_j) \end{aligned}$$

## ML-EM for BMM

Input: Data  $x_1, \dots, x_n, x \in \mathbb{R}^d$ . Number of clusters  $K$ .

Output: BMM parameters  $\pi, \theta$  and cluster assignment distributions  $\phi_i$ .

1. Initialize  $\pi^{(0)}$  and  $\theta^{(0)}$  in some way.

2. At iteration  $t$ ,

E-Step: For  $i=1, \dots, n$  and  $j=1, \dots, K$  set.

$$\phi_i^{(t)}(j) = \frac{\pi_j \text{Binomial}(x_i | \theta_j)}{\sum_{k=1}^K \pi_k \text{Binomial}(x_i | \theta_k)}$$

M-Step: Set

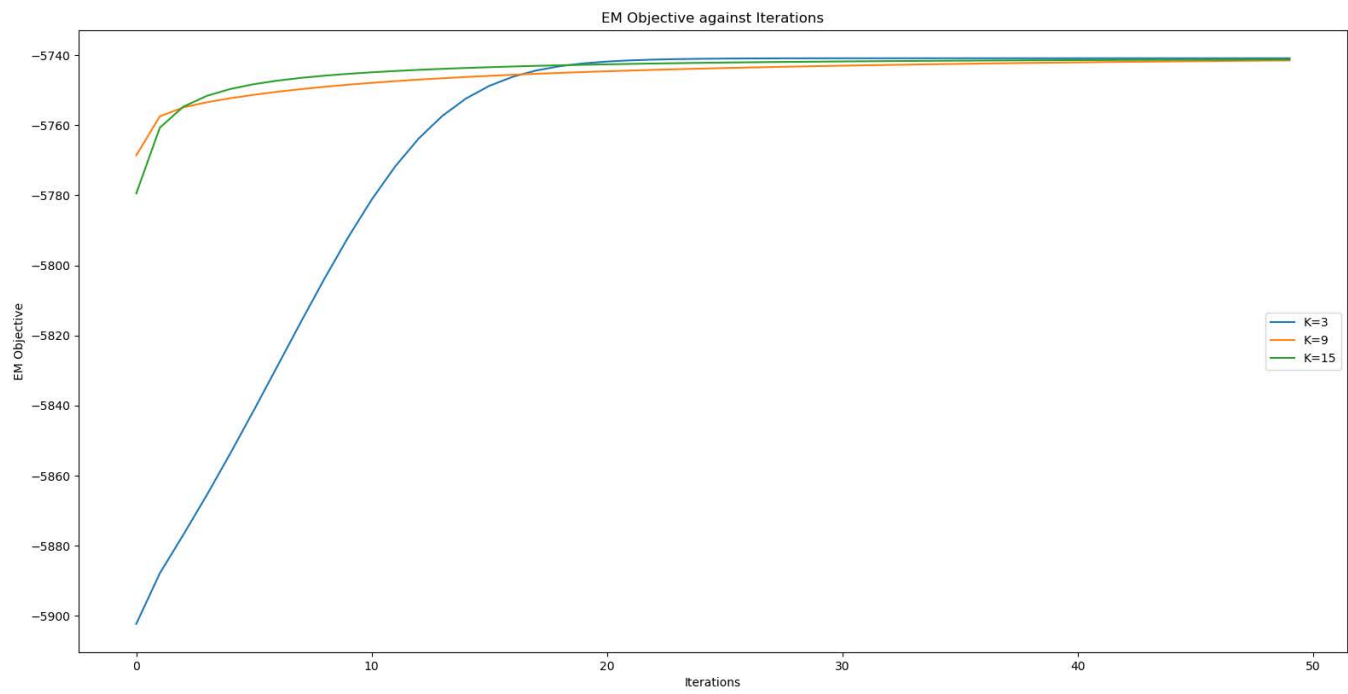
$$n_j^{(t)} = \sum_{i=1}^n \phi_i^{(t)}(j)$$

$$\theta_j = \frac{\sum_{i=1}^n \phi_i^{(t)}(j) x_i}{\sum_{i=1}^n \phi_i^{(t)}(j)}$$

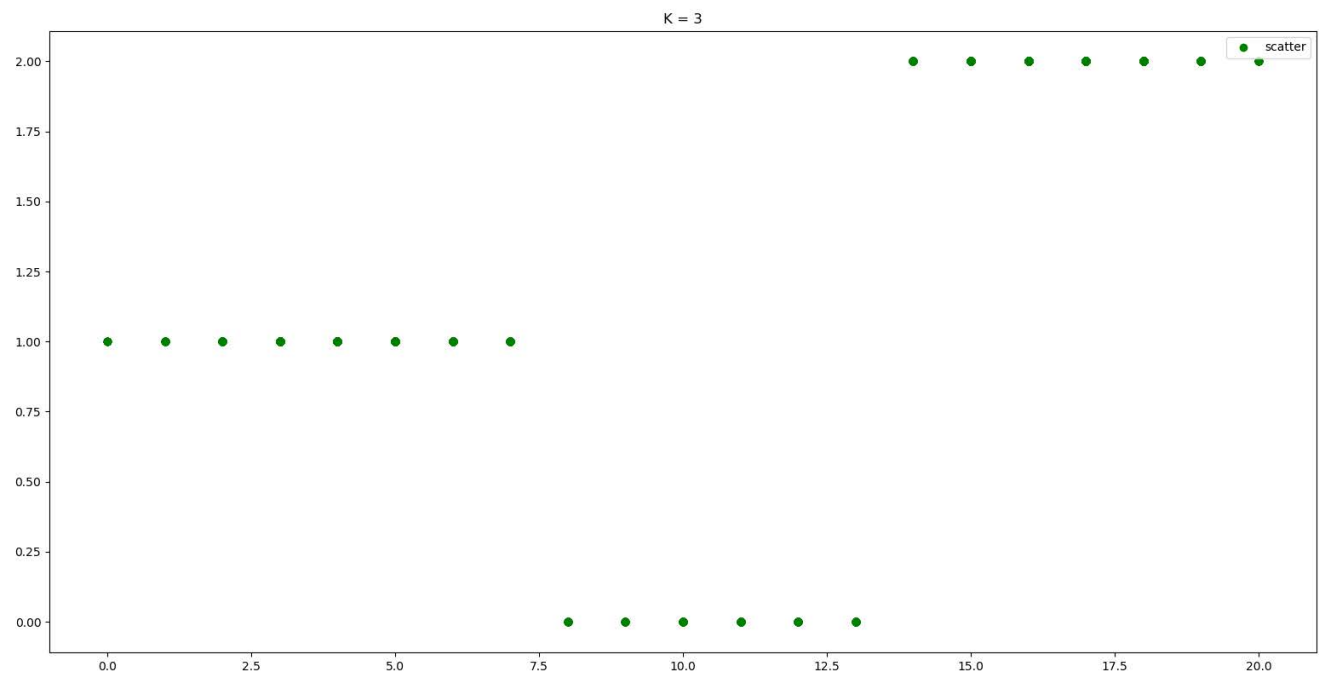
$$\pi_j = \frac{n_j^{(t)}}{n}$$

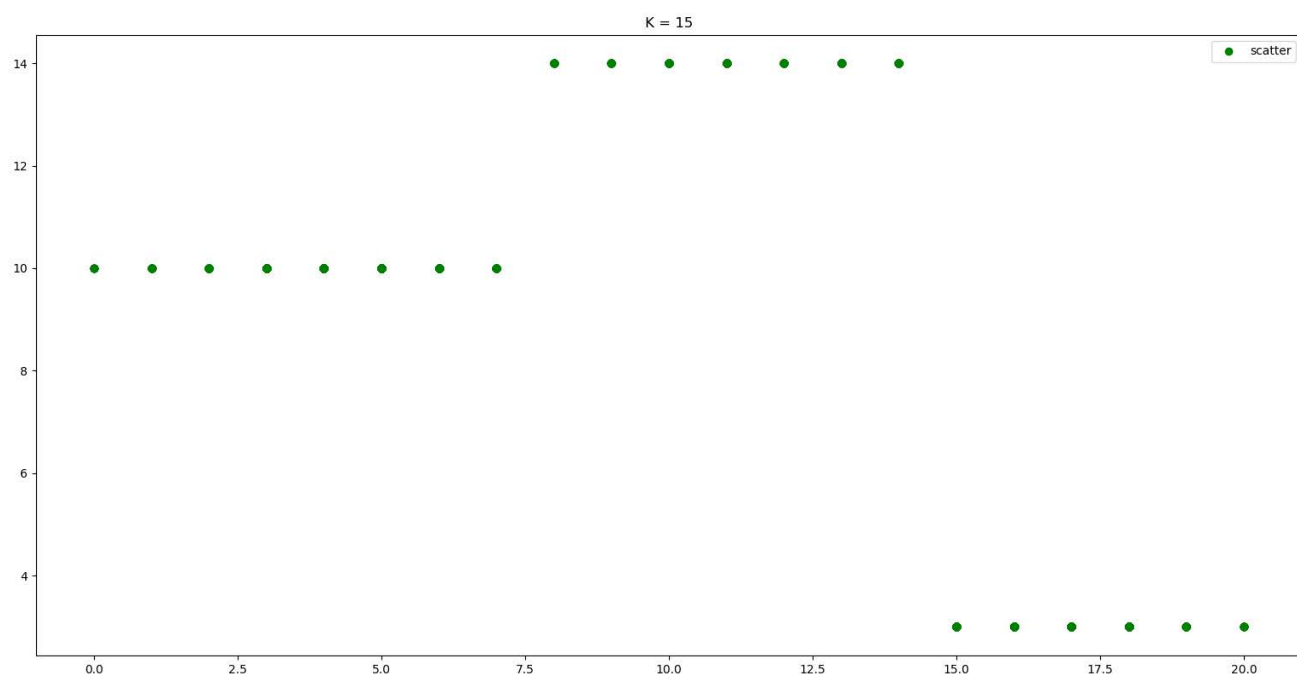
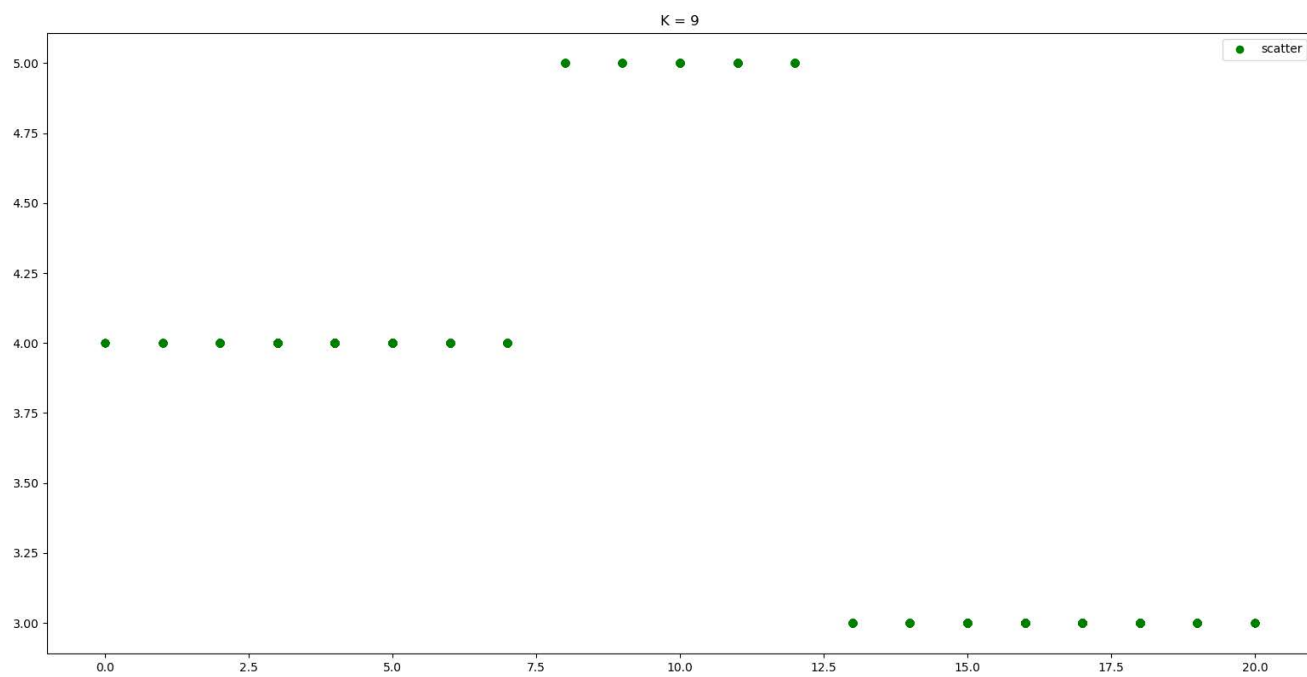
3. Calculate  $f_t = \ln p(x | \pi, \theta)$

b.



c.





Problem 2.

$$p(x, G, \pi, \theta) = \prod_{i=1}^n p(x_i, G_i, \pi, \theta) = \prod_{i=1}^n p(x_i | G_i, \pi, \theta) p(G_i | \pi) \prod_{j=1}^K p(\theta_j) p(\pi_j)$$

$$\begin{aligned} q(G_i=j) &= q(G_i=j) \propto e^{\mathbb{E}[\ln p(x, G_i=j | \pi, \theta_j)]} \\ &\propto e^{\mathbb{E}[\ln p(x | \theta_j)] + \mathbb{E}[\ln p(G_i=j | \pi)]} \\ &\propto e^{x_i \mathbb{E}[\ln \theta_j] + (20-x_i) \mathbb{E}[\ln (1-\theta_j)] + \mathbb{E}[\ln \pi_j]} \\ &\propto e^{x_i [\psi(\alpha_j') - \psi(\alpha_j' + \beta_j')] + (20-x_i) [\psi(\beta_j') - \psi(\alpha_j' + \beta_j')] + \psi(\alpha_j') - \psi(21\alpha_j')} \end{aligned}$$

$$\begin{aligned} q(\pi) &= q(\pi) \propto e^{\sum_{i=1}^n \mathbb{E}[\ln p(G_i=j | \pi)] + \ln p(\pi)} \\ &\propto e^{\sum_{i=1}^n \sum_{j=1}^K \phi_i(j) \ln \pi_j + \sum_{j=1}^K (\alpha-1) \ln \pi_j} \\ &\propto \prod_{j=1}^K \pi_j^{\sum_{i=1}^n \phi_i(j) + \alpha - 1} \\ &= \text{Dirichlet}(\alpha') \end{aligned}$$

where  $\alpha_j' = n_j + \alpha_j$ ,  $n_j = \sum_{i=1}^n \phi_i(j)$ .

$$\begin{aligned} q(\theta_j) &\propto e^{\sum_{i=1}^n \mathbb{E}[\ln p(x | G_i=j, \pi, \theta)] + \ln p(\theta_j)} \\ &\propto e^{[\sum_{i=1}^n \phi_i(j) x_i + \alpha - 1] \ln \theta_j + [\sum_{i=1}^n \phi_i(j) (20-x_i) + \beta - 1] \ln (1-\theta_j)} \\ &= \text{Beta}(\alpha', \beta') \end{aligned}$$

where  $\alpha' = \sum_{i=1}^n \phi_i(j) x_i + \alpha$ ,  $\beta' = \sum_{i=1}^n \phi_i(j) (20-x_i) + \beta$ .

$$\begin{aligned}
\mathcal{L} &= \mathbb{E}[\ln p(x, c, \pi, \theta)] - \mathbb{E}[\ln q] \\
&= \sum_{i=1}^n \sum_{j=1}^K \mathbb{E}[\mathbb{1}(C_i=j) [\ln p(x_i|\theta_j) + \ln \pi_j]] + \sum_{j=1}^K \mathbb{E}[\ln p(\theta_j)] + \mathbb{E}[\ln p(\pi)] \\
&\quad - \sum_{i=1}^n \mathbb{E}[\ln q(C_i)] - \sum_{k=1}^K \mathbb{E}[\ln q(\theta_k)] - \mathbb{E}[\ln q(\pi)]
\end{aligned}$$

$$\begin{aligned}
\mathbb{E}[\mathbb{1}(C_i=j) [\ln p(x_i|\theta_j) + \ln \pi_j]] &= \phi_i(j) \mathbb{E}[\ln p(x_i|\theta_j) + \ln \pi_j] \\
&= \phi_i(j) \{ \mathbb{E}[\ln p(x_i|\theta_j)] + \mathbb{E}[\ln \pi_j] \} \\
&= \phi_i(j) \{ \ln(\bar{x}_i) + x_i [\psi(a_j') - \psi(a_j' + b_j')] + (20 - x_i) [\psi(b_j') - \psi(a_j' + b_j')] + \psi(\alpha_j') - \psi(\sum_k \alpha_k') \}
\end{aligned}$$

$$\mathbb{E}[\ln p(\theta_j)] = (a_j' - 1) [\psi(a_j') - \psi(a_j' + b_j')] + (b_j' - 1) [\psi(b_j') - \psi(a_j' + b_j')] - \ln B(a_j', b_j')$$

$$\mathbb{E}[\ln p(\pi)] = \sum_i (\alpha_i' - 1) [\psi(\alpha_i') - \psi(\sum_k \alpha_k')] - \ln B(\alpha_j').$$

$$\mathbb{E}[\ln q(C_i)] = \sum_{j=1}^K \phi_i(j) \cdot \ln \phi_i(j)$$

$$\mathbb{E}[\ln q(\theta_j)] = (a_j' - 1) [\psi(a_j') - \psi(a_j' + b_j')] + (b_j' - 1) [\psi(b_j') - \psi(a_j' + b_j')] - \ln B(a_j', b_j')$$

$$\mathbb{E}[\ln q(\pi_j)] = \sum_i (\alpha_i' - 1) [\psi(\alpha_i') - \psi(\sum_k \alpha_k')] - \ln B(\alpha_j').$$

where  $a_j' = a_j^{(t)}$ ,  $b_j' = b_j^{(t)}$ ,  $\alpha_j' = \alpha_j^{(t)}$ .

VI for BMM.

Input: Data  $x_1, \dots, x_n, x \in \mathbb{R}^d$ . Number of cluster  $K$

Output: Parameters for  $q(\lambda), q(\theta_j), q(c_i)$ .

1. Initialize  $(\alpha_1^{(0)}, \dots, \alpha_K^{(0)}), (a_j^{(0)}, b_j^{(0)})$  in some way.

2. At iteration  $t$ ,

(a) Update  $q(c_i)$  for  $i=1, \dots, n$ :

$$\phi_i^{(t)}(j) = \frac{e^{\lambda_i t_1(j) + (20 - \lambda_i) t_2(j) + t_3(j)}}{\sum_{k=1}^K e^{\lambda_i t_1(k) + (20 - \lambda_i) t_2(k) + t_3(k)}}$$

where  $t_1(j) = \psi(a_j^{(t-1)}) - \psi(a_j^{(t-1)} + b_j^{(t-1)})$   $t_2 = \psi(b_j^{(t-1)}) - \psi(a_j^{(t-1)} + b_j^{(t-1)})$   
 $t_3 = \psi(\alpha_j^{(t-1)}) - \psi(\sum_K \alpha_K^{(t-1)})$

(b) Set  $n_j^{(t)} = \sum_{i=1}^n \phi_i^{(t)}(j)$  for  $j=1, \dots, K$ .

(c) Update  $q(\lambda)$  by setting:

$$\alpha_j^{(t)} = \alpha + n_j^{(t)}.$$

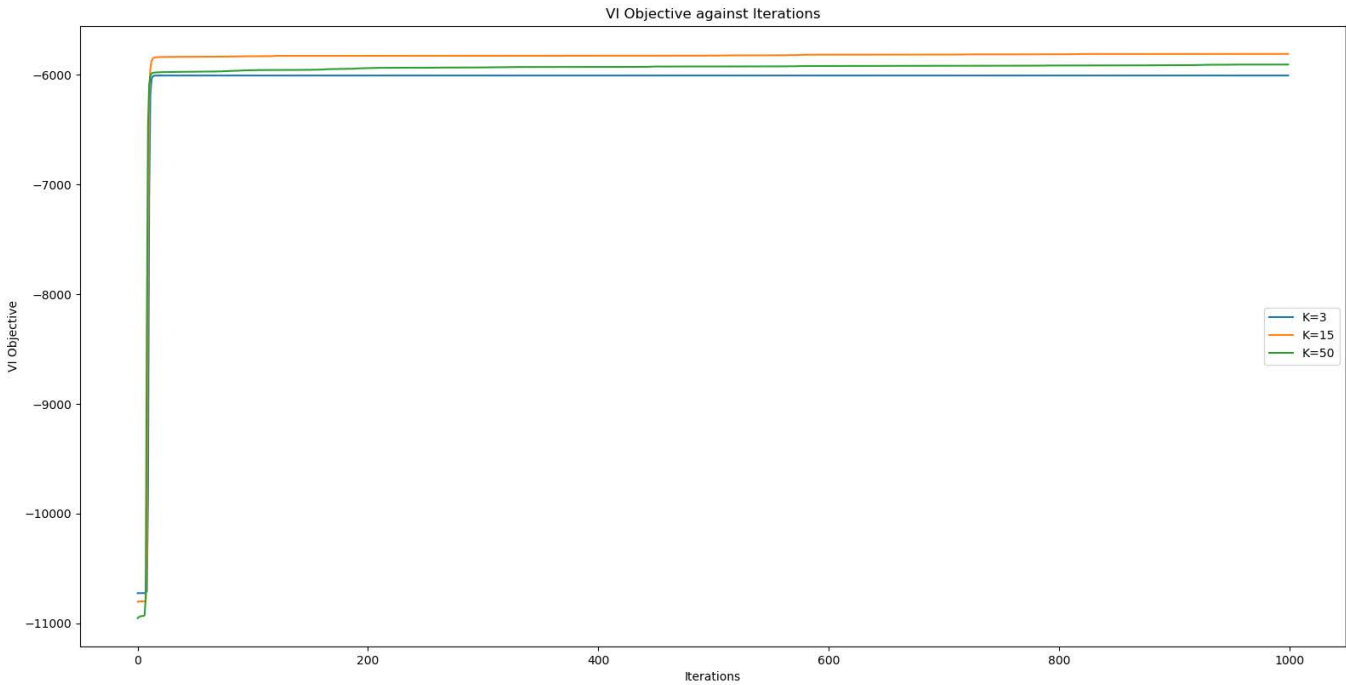
for  $j=1, \dots, K$

(d). Update  $q(\theta_j)$  for  $j=1, \dots, K$  by setting:

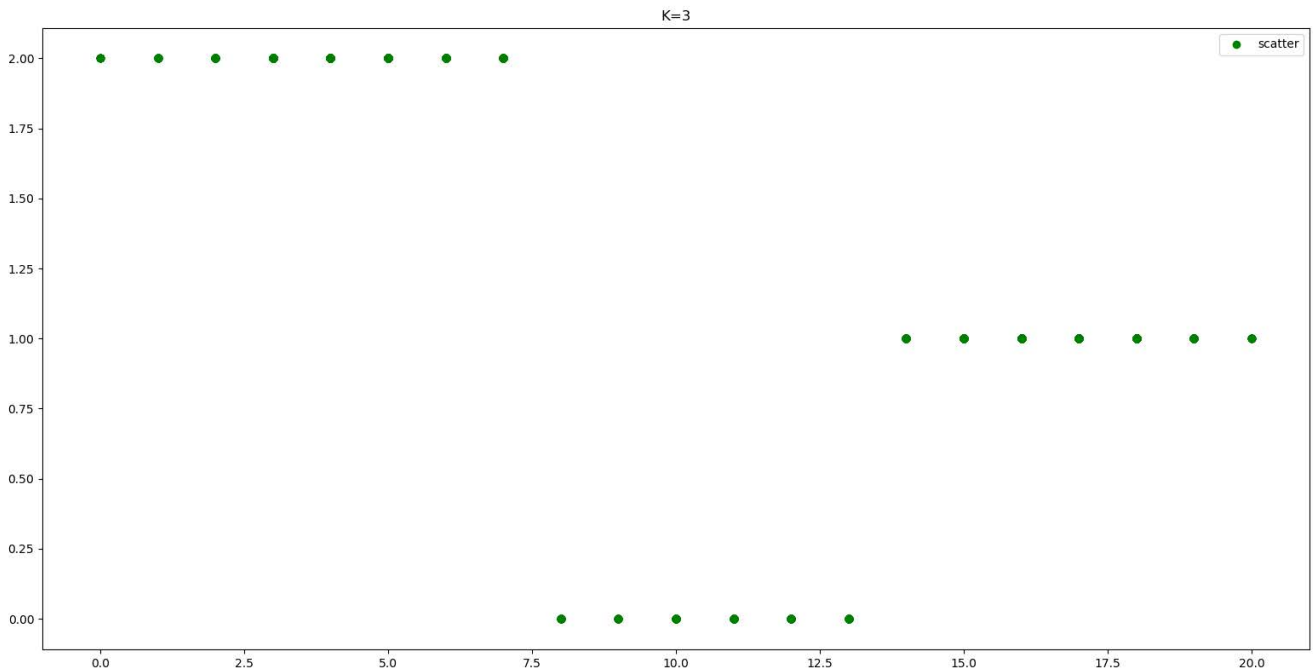
$$\cancel{\alpha^{(t+1)}} = a^{(t)} = \frac{\sum_{i=1}^n \phi_i^{(t)}(j)}{\sum_{i=1}^n \phi_i^{(t)}(j)} x_i + a, \quad b^{(t)} = \frac{\sum_{i=1}^n \phi_i^{(t)}(j)}{\sum_{i=1}^n \phi_i^{(t)}(j)} (20 - x_i) + b.$$

(e) Calculate  $\mathcal{L} = \mathbb{E}[\ln p(x, \lambda, c, \theta)] - \mathbb{E}[\ln q]$

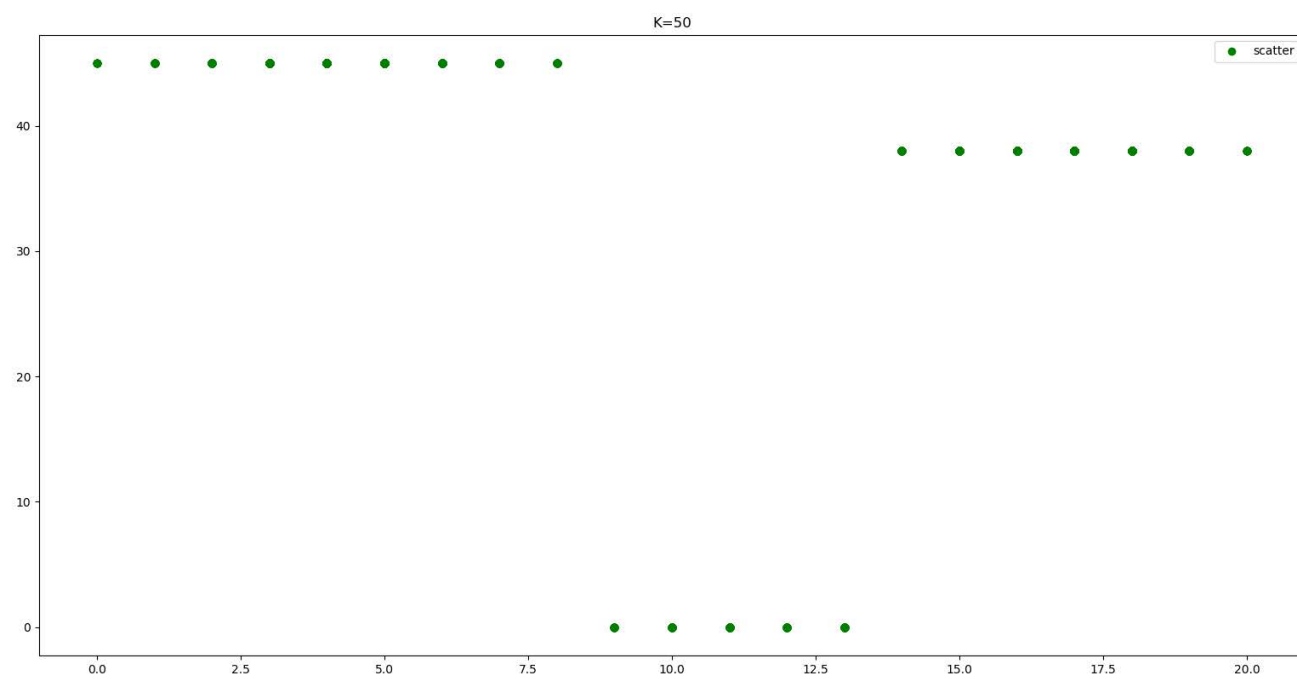
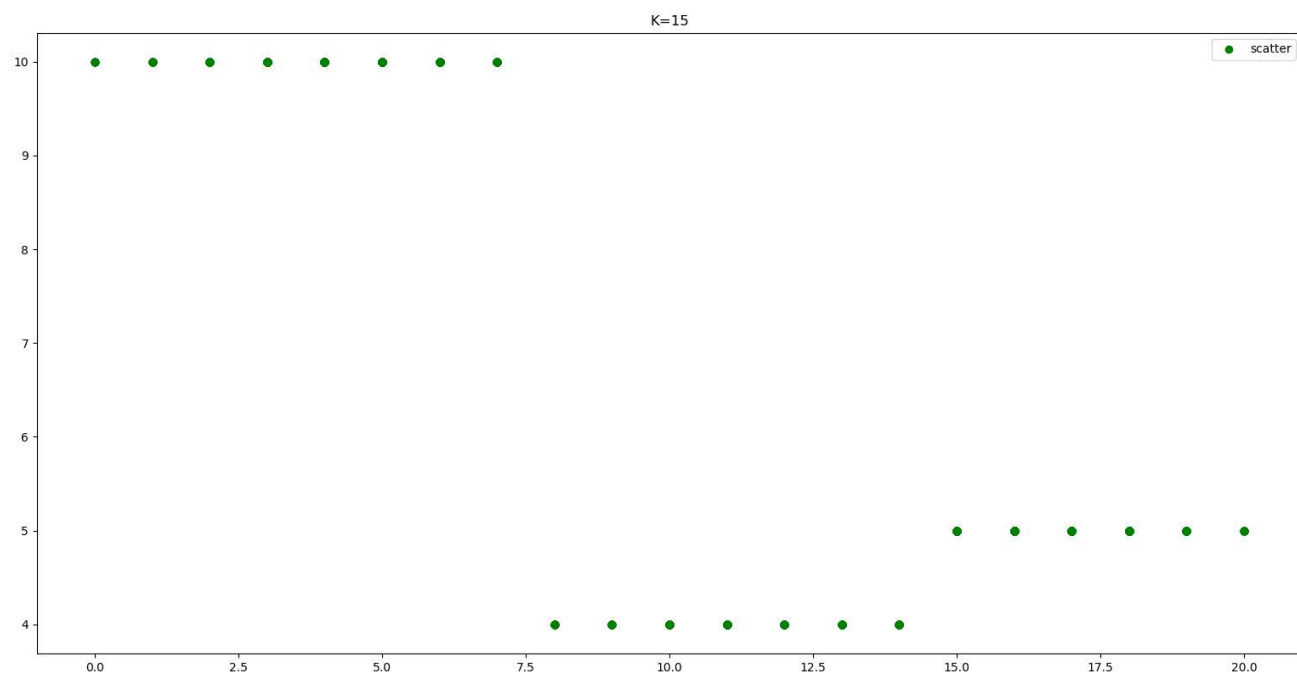
b.



c.







# Problem 3.

$$P(\theta | \{x_i: G=j\}) \propto \prod_{i=1}^n p(x_i | \theta_j)^{\mathbb{1}(G=j)} p(\theta_j).$$

$$\propto \theta_j^{\sum_{i=1}^n \mathbb{1}(G=j) x_i} (1-\theta_j)^{\sum_{i=1}^n \mathbb{1}(G=j) (20-x_i)} \theta_j^{a-1} (1-\theta_j)^{b-1}$$

$$\propto \theta_j^{a + \sum_{i=1}^n \mathbb{1}(G=j) x_i - 1} (1-\theta_j)^{b + \sum_{i=1}^n \mathbb{1}(G=j) (20-x_i) - 1}.$$

$$= \text{Beta}(a + \sum_{i=1}^n \mathbb{1}(G=j) x_i, b + \sum_{i=1}^n \mathbb{1}(G=j) (20-x_i)).$$

$$P(G=j | x_i, \theta, G_i) \propto p(x_i | \theta_j) \frac{n_j^{(-i)}}{\alpha + n - 1}.$$

$$\propto \text{Binomial}(x_i | 20, \theta_j) \cdot \frac{n_j^{(-i)}}{\alpha + n - 1}.$$

$$P(G=\text{new} | x_i, \theta, G_i) \propto \frac{\alpha}{\alpha + n - 1} \int p(x_i | \theta) p(\theta) d\theta$$

Beta-Binomial distribution.

$$\propto \frac{\alpha}{\alpha + n - 1} \binom{20}{x_i} \frac{B(a+x_i, b+20-x_i)}{B(a, b)}.$$

Dirichlet process mixture model:

1. ~~Iter~~ Initialize in same way.
2. At iteration  $t$ , re-index clusters. Sample all variables below using the most recent values of the other values.

1. For  $i=1, \dots, n$ .

(a) For all  $j$  such that  $n_j^{(-i)} > 0$ , set:

$$\hat{\phi}_i(j) = p(x_i | \theta_j) n_j^{(-i)} / (\alpha + n - 1) = \frac{n_j^{(-i)}}{\alpha + n - 1} \cdot \text{Binomial}(x_i | 20, \theta_j).$$

(b) For a new value  $j'$ , set.

$$\hat{\phi}_i(j') = \frac{\alpha}{\alpha + n - 1} \binom{20}{x_i} \frac{B(a+x_i, b+20-x_i)}{B(a, b)}.$$

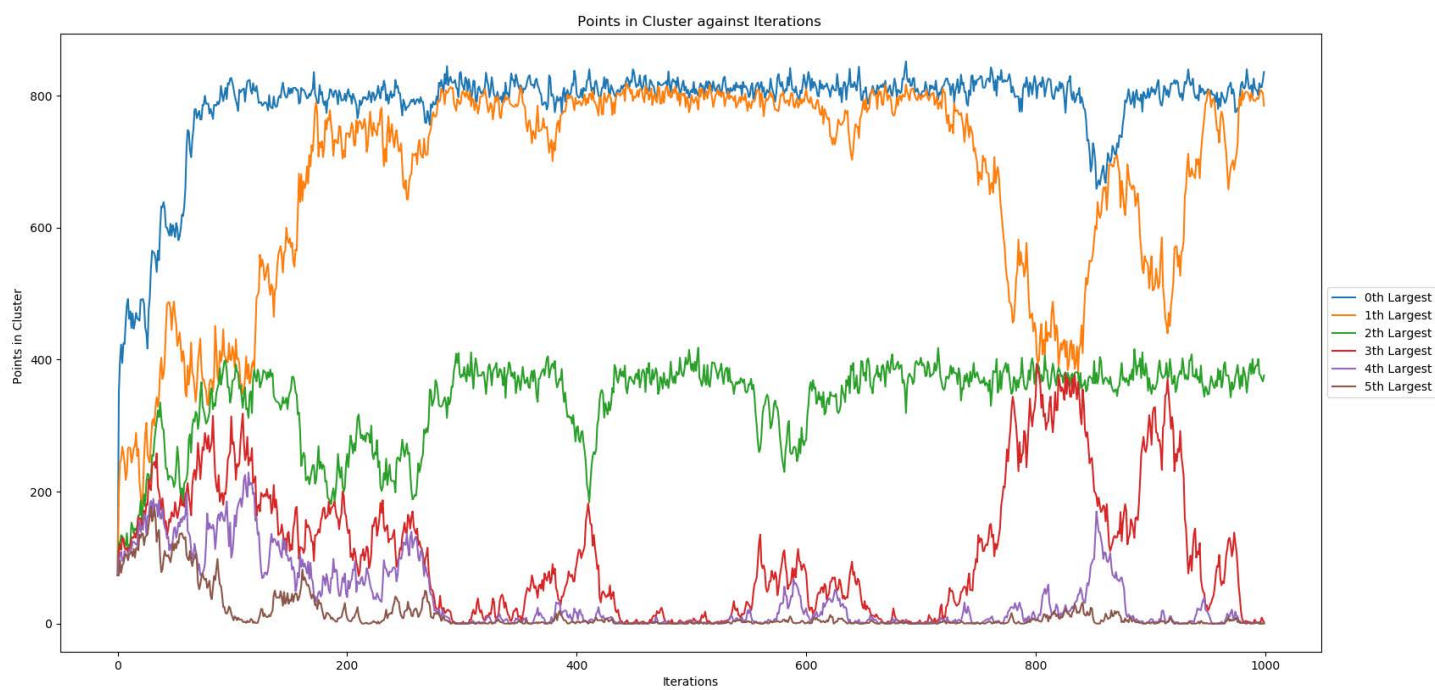
(c). Normalize  $\hat{\phi}_i$  and sample the index  $G_i$  from a discrete distribution.

(d) If  $G_i = j'$ , generate  $\theta_{j'} \sim \text{Beta}(a+x_i, b+20-x_i)$ .

2. For  $j=1, \dots, K^{(t)}$  generate

$$\theta_j \sim \text{Beta}(a + \sum_{i=1}^n \mathbb{1}(G_i=j) x_i, b + \sum_{i=1}^n \mathbb{1}(G_i=j) (20-x_i)).$$

b.



c.

