

P1.  $y_i \sim \text{Bern}(\pi)$ ,  $x_{i,d} | y_i \sim \text{Pois}(\lambda_{y_i,d})$   $\lambda_{y,d} \sim \text{Gamma}(2, 1)$

$$(a) \arg \max_{\lambda, \pi} \sum_{i=1}^n \ln p(y_i | \pi) + \sum_{d=1}^D (\ln p(\lambda_{0,d}) + \ln p(\lambda_{1,d}) + \sum_{i=1}^n \ln p(x_{i,d} | \lambda_{y_i,d}))$$

Only first part depends on  $\pi$ .

$$\therefore \hat{\pi} = \arg \max_{\pi} \sum_{i=1}^n \ln p(y_i | \pi) \\ = \arg \max_{\pi} \sum_{i=1}^n \ln(\pi^{y_i} (1-\pi)^{1-y_i}) = \arg \max_{\pi} \sum_{i=1}^n (y_i \ln \pi + (1-y_i) \ln(1-\pi))$$

By taking the derivative and setting to zero:

$$\sum_{i=1}^n \left( \frac{1}{\pi} y_i + \frac{y_i - 1}{1-\pi} \right) = \sum_{i=1}^n \left( \frac{y_i}{\pi} + \frac{y_i - 1}{1-\pi} \right) = 0 \Rightarrow \sum_{i=1}^n \left( \frac{y_i - \pi}{\pi(1-\pi)} \right) = 0 \\ \hat{\pi} = \frac{\sum_{i=1}^n y_i}{n}$$

$$(b) \sum_{d=1}^D \left\{ \ln \lambda_{0,d} e^{-\lambda_{0,d}} + \ln \lambda_{1,d} e^{-\lambda_{1,d}} + \sum_{i=1}^n \left[ (1-y_i) \ln \frac{\lambda_{0,d}^{x_i} e^{-\lambda_{0,d}}}{x_i!} + y_i \ln \frac{\lambda_{1,d}^{x_i} e^{-\lambda_{1,d}}}{x_i!} \right] \right\}$$

① MLE for  $\lambda_0$

$$\text{By taking derivative:} \quad \left[ \frac{1}{\lambda_{0,d}} - 1 + \sum_{i=1}^n \left( \frac{x_i}{\lambda_{0,d}} - 1 \right) (1-y_i) \right] = 0 \\ \left[ 1 - \lambda_{0,d} + \sum_{i=1}^n (x_{i,d} - \lambda_{0,d}) (1-y_i) \right] = 0 \\ \left[ 1 - \lambda_{0,d} + \sum_{i=1}^n (1-y_i) x_{i,d} - \sum_{i=1}^n (1-y_i) \lambda_{0,d} \right] = 0 \\ 1 - \lambda_{0,d} + \sum_{i=1}^n (1-y_i) x_{i,d} - \sum_{i=1}^n (1-y_i) \lambda_{0,d} = 0 \\ \hat{\lambda}_{0,d} = \frac{1 + \sum_{i=1}^n (1-y_i) x_{i,d}}{1 + \sum_{i=1}^n (1-y_i)}$$

② MLE for  $\lambda_1$

$$\text{By taking derivative:} \quad \left[ \frac{1}{\lambda_{1,d}} - 1 + \sum_{i=1}^n y_i (x_{i,d} - \lambda_{1,d}) \right] = 0 \\ \frac{1}{\lambda_{1,d}} - 1 + \sum_{i=1}^n y_i x_{i,d} - \sum_{i=1}^n y_i \lambda_{1,d} = 0 \\ \hat{\lambda}_{1,d} = \frac{1 + \sum_{i=1}^n y_i x_{i,d}}{1 + \sum_{i=1}^n y_i}$$

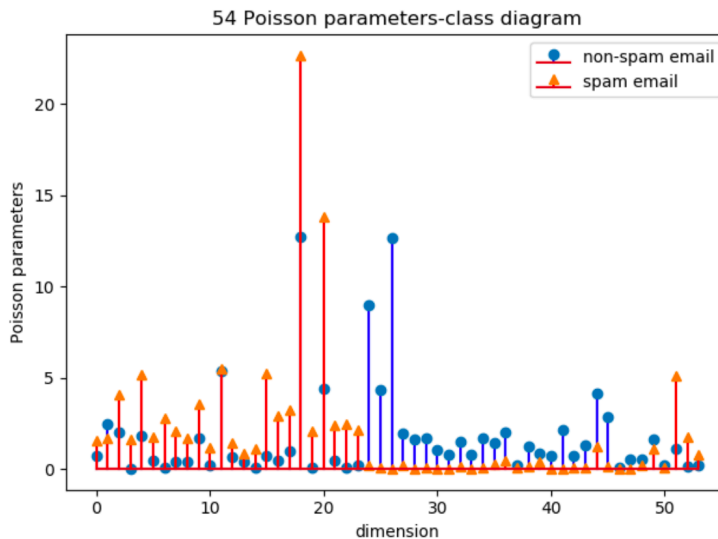
P2.

(a)

True\Pred	1	0
1	1712	101
0	491	2296

$$\text{Accuracy} = 87.13\%$$

(b)

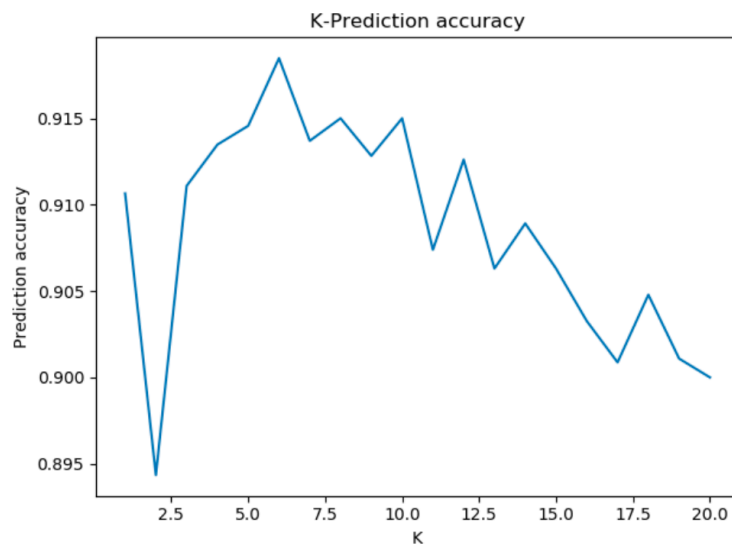


Dimension 16 : "free", The poisson parameter for spam email is much higher than that for non-spam email. The poisson parameter for spam email is around 5.

Dimension 52 : " ! ", The poisson parameter for spam email is higher than that for non-spam email. It's around 2 or 3.

These 2 dimensions' parameters are not the highest one. It seems that "free" and "!" more likely shows in spam email. This is in line with our common sense. Spam emails always use "free" "!" to attract people.

(c) KNN's result.



P3.

a) shown as code.

b)

	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0
5	1.9662770832011400	1.9331367430388900	1.9234218492955000	1.9221994122141700	1.9247710572433900	1.929214584821610	1.9346361805691700	1.9405852698053700	1.9468221527738300	1.9532147535652100
7	1.9201644559708700	1.9048781595333700	1.9080822266835800	1.9159037429601700	1.9248062582061500	1.9337036635460000	1.942256089136370	1.9503824005884000	1.9580954559045700	1.9654404467105400
9	1.8976501853604700	1.902521011958540	1.9176498037931700	1.9325165811150100	1.9457018068985400	1.9572371755057400	1.9674056705298200	1.9764941762129700	1.9847432013773700	1.9923436357077300
11	1.8905087802491300	1.9149830815209600	1.938851035693020	1.9579385936670300	1.973218217700310	1.9857666177104800	1.9963775483262600	2.0056056170174600	2.013837822849690	2.0213471533170600
13	1.895850359906790	1.9355881004200100	1.9645996488381400	1.9855043698573200	2.0013166575965500	2.013880822760730	2.024312725963830	2.0333090940990900	2.041319769316060	2.048643790039850
15	1.9096052208844500	1.959551103489170	1.9908059319304000	2.0119178086292600	2.027372618959860	2.039467464315050	2.0494656331464800	2.058107096465770	2.065847425140800	2.072978163729000

(c) The minimum RMSE is 1.8905 with  $b=1$ ,  $\sigma^2=0.1$  This result

is better than what we did in homework 1. In homework 1, the RMSE is between around 2.65 to 2.85. In this method, the RMSE is around 1.89 to 2.07, which is lower than that in homework 1. In this case, Gaussian process is better.

The drawback is that GP's computation time. In practice it's hard to work with more than a few thousand points. The cost of GP computation is very high.

(d)

