

Homework Assignment N°2

AML3

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Contents

1	Exercise 1	3
1.1	Part a	3
1.2	Part b	3
2	Exercise 2	4
2.1	Part a	4

1 Exercise 1

1.1 Part a

We want to compute for each sample in the data the responsibility $\gamma(z_k)$ of each Bernoulli distribution.

What we call responsibility is the probability knowing the data occurred that model z_k is involved. We can rewrite it in term of probabilities:

$$\gamma(z_k) = P(z_k|x) = \frac{p(x|z_k)p(z_k)}{p(x)}$$

Where $p(z_k)$ is the prior of each model, $p(x|z_k)$ is given by the distribution z_k (here Bernoulli) and $p(x)$ is the sum over all models in the mixture of the probability they produced x .

For example with the first point:

- The prior is given by our assumptions: $p(z_k) = [0.5, 0.5]$,
- Each model is a Bernoulli distribution,
hence $p(x|z_k) = \binom{x_0 + x_1}{x_0} \mu_k^{x_0} (1 - \mu_k)^{x_1}$
- And the total probability of x is just the marginalization over all models:
 $p(x) = \sum_{z_k} p(x|z_k)p(z_k)$

We finally obtain the following values:

$$p((1, 4)) \approx 0.243$$

and the responsibilities:

$$\gamma(z_1) = 0.678$$

$$\gamma(z_2) = 0.322$$

We repeat the same procedure for all given inputs and obtain the following results:

x	$\gamma(z_{n1})$	$\gamma(z_{n2})$
(1, 4)	0.678	0.322
(3, 2)	0.345	0.655
(4, 1)	0.208	0.792
(2, 3)	0.513	0.487

1.2 Part b

We now want to improve our mixture of model in order to better explain the data we saw. We must compute a new value for the prior and the parameter of each Bernoulli distribution.

Using the responsibilities we just computed in the previous step, the EM algorithm gives us formulas to update those values:

The new prior of model z_k :

$$\pi_k^2 = \frac{\gamma(z_k)}{\sum_i \gamma(z_i)}$$

The new proportion of head of model z_k :

$$\mu_k^2 = \frac{\gamma(z_k)\mu_k}{\sum_i \gamma(z_i)\mu_i}$$

But because we now operate in batch after seeing multiple data, we need to modify those equations by summing over all seen points:

$$\pi_k^2 = \frac{\sum_n \gamma(z_{nk})}{\sum_n \sum_i \gamma(z_{ni})}$$

$$\mu_k^2 = \frac{\sum_n \gamma(z_{nk})\mu_k}{\sum_n \sum_i \gamma(z_{ni})\mu_i}$$

With the responsibilities from the previous exercise and the parameters given, we obtain the following results:

$$\pi^2 = [0.436, 0.564]$$

$$\mu^2 = [0.409, 0.570]$$

This means that we have two models with priors π^2 . Each one of them is still a Bernoulli model. The first one gives a probability of 0.409 to H and the second a probability of 0.570 to H.

2 Exercise 2

2.1 Part a