



[Unit 4 Unsupervised Learning \(2](#)

[Course](#) > [weeks](#))

> [Lecture 15. Generative Models](#) >

7. Prediction

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## 7. Prediction

## Prediction



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## Predictions of a Generative Multinomial Model

1/1 point (graded)

Consider using a multinomial generative model  $M$  for the task of binary classification consisting of two classes which are denoted by + (positive class) and - (negative class).

Let the parameters of  $M$  that maximize the likelihood of training data for the positive class be denoted by  $\theta^+$  and for the negative class be denoted by  $\theta^-$ .

Also, suppose that we classify a new document  $D$  to belong to the positive class iff

$$\log \frac{P(D|\theta^+)}{P(D|\theta^-)} \geq 0$$

where  $P(D|\theta)$  stands for the probability that document  $D$  is generated using a multinomial distribution with parameters  $\theta$ .

Which of the following option(s) is/are true about this generative classifier? Choose all that apply from the statements below:

☒ A document is classified as positive iff  $P(D|\theta^+) \geq P(D|\theta^-)$

☐ A document is classified as positive iff  $P(D|\theta^+) < P(D|\theta^-)$

☒ The generative classifier  $M$  can be shown to be equivalent to a linear classifier given by  $\sum_{w \in W} (\text{count}(w) \theta'_w) \geq 0$  where  $\theta'_w = \log \frac{\theta_w^+}{\theta_w^-}$

☐ The generative classifier  $M$  can be shown to be equivalent to a linear classifier given by  $\sum_{w \in W} (\text{count}(w) \theta'_w) \geq 0$  where  $\theta'_w = \log \frac{\theta_w^-}{\theta_w^+}$



**Solution:**

Note that we classify a new document  $D$  to belong to the positive class iff

$$\log \left( \frac{P(D|\theta^+)}{P(D|\theta^-)} \right) \geq 0 \text{ and to the negative class otherwise.}$$

$$\log \frac{P(D|\theta^+)}{P(D|\theta^-)} \geq 0$$

is equivalent to

$$P(D|\theta^+) \geq P(D|\theta^-)$$

.  
Recall from the lecture that,

$$\log \frac{P(D|\theta^+)}{P(D|\theta^-)}$$

$$= \log P(D|\theta^+) - \log P(D|\theta^-)$$

$$= \log \prod_{w \in W} (\theta_w^+)^{\text{count}(w)} - \log \prod_{w \in W} (\theta_w^-)^{\text{count}(w)}$$

$$= \sum_{w \in W} \text{count}(w) \log \theta_w^+ - \sum_{w \in W} \text{count}(w) \log \theta_w^-$$

$$= \sum_{w \in W} \text{count}(w) \log \frac{\theta_w^+}{\theta_w^-}$$

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You have used 1 of 2 attempts

**i** Answers are displayed within the problem

## Linear Classifier of the Generative Multinomial Model

1/1 point (graded)

Consider the prediction classifier for the two classes  $\theta^+$  and  $\theta^-$  introduced in the above video. For this problem, let 0 and 1 represent the classes + and -, respectively.

Let  $W = \{\text{Thor}, \text{Loki}, \text{Hulk}\}$ . Let

$p(\text{Thor}|0) = p(\text{Loki}|0) = p(\text{Hulk}|0) = 1/3$  and let  
 $p(\text{Thor}|1) = p(\text{Loki}|1) = 1/4$  and  $p(\text{Hulk}|1) = 1/2$ .

We see the following document  $D = \text{Thor Thor Hulk Loki Loki}$ . To what class would you classify the document to using the linear classifier for the generative multinomial model? (Type "0" for class 0 (+) and "1" for class 1 (-)).

✓ Answer: 0

### Solution:

The counts of the words are

$\text{count}(\text{Thor}) = 2, \text{count}(\text{Loki}) = 2, \text{count}(\text{Hulk}) = 1$ .

Using the notation developed in the video,

$$\hat{\theta}_{\text{Thor}} = \log \left( \frac{4}{3} \right) \approx 0.124939,$$

$$\hat{\theta}_{\text{Loki}} = \log\left(\frac{4}{3}\right) \approx 0.124939,$$

$$\hat{\theta}_{\text{Hulk}} = \log\left(\frac{2}{3}\right) \approx -0.176091.$$

Therefore,

$$\sum_{w \in \mathcal{W}} \text{count}(w) \hat{\theta}_w = 4 \cdot 0.124939 - 0.176091 > 0,$$

which would classify the document to class 0 (+).

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You have used 1 of 1 attempt

**i** Answers are displayed within the problem

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|----------|--|---|
| <b>?</b> | <u>How do we get the different parameters for positive and negative cases when we don't have labels to begin with?</u>                           | 1 |
| <b>✓</b> | <u>Should we use natural log.(ln) or log10 in the second question?</u><br><u>Should we use natural log.(ln) or log10 in the second question?</u> | 7 |

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