
Generative Models Revisited

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Which are generative models?

$p(y|x; \theta)$ a) Yes
 b) No

$p(x; \theta)$

$p(x,y; \theta)$

Generative Models

hidden

$$s \sim p(k)$$

$$s \sim p(y)$$

$$s \sim N(0, I)$$

$$s \sim N(0, I)$$

observed

$$p(\mathbf{x}|s=k) = N(\boldsymbol{\mu}_k, \boldsymbol{\Sigma}_k)$$

$$p(\mathbf{x}|s=y) = N(\boldsymbol{\mu}_y, \boldsymbol{\Sigma}_y) = \prod_j N(\mu_{jy}, \sigma_{jy}^2)$$

$$\mathbf{x} = \mathbf{A} \mathbf{s} + \boldsymbol{\varepsilon}$$

$$\mathbf{x} = \mathbf{A} \mathbf{s} + \boldsymbol{\varepsilon} \quad \mathbf{y} = \mathbf{B} \mathbf{s} + \boldsymbol{\varepsilon}$$

Dimensions:

\mathbf{s} is $k \times 1$

\mathbf{x} is $p \times 1$

\mathbf{y} is $q \times 1$

A) PCA

B) CCA

C) GMM

D) Naïve Bayes

Generative models for classification

- **Naïve Bayes**
- **K-Means/GMM**

Generative models for classification

- **K-Means for prediction**

- Cluster training points using k-means
- Find the most frequent label (or average of real-valued labels) for the points in each cluster

$$\hat{y}_k = \sum_i \sum_k I(x_i \text{ in cluster } k) y_i$$

- To predict at a new x
 - find which cluster centroid μ_k the new x is closest to
 - look up the label or \hat{y}_k for that cluster k

Generative models for classification

- **GMM for prediction**

- **Estimate GMM**

- estimate π_k, μ_k, Σ_k for each cluster

- **Find the expected value of the label for each cluster**

- weighting by the degree of membership of x_i in the cluster k

$$\hat{y}_k = \sum_i \sum_k p(\text{cluster}=k \mid x=x_i) y_i$$

- **To predict at a new x**

- find the probability of x belonging to each cluster

$$p(\text{cluster}=k \mid x=x_i)$$

- look up the \hat{y}_k for that cluster

- the prediction is a weighted combination of those cluster labels

$$\sum_k p(\text{cluster}=k \mid x=x_i) \hat{y}_k$$

This is not common, so I just made it up. Is there a better way?

LDA – Review

- Which symbol corresponds to each of these
 - $P(\text{topic}=k)$
 - $P(\text{topic}=k)$ in document d
 - $P(\text{word}=w_j \mid \text{topic}=k)$

A) β_{jk}

B) θ_k

C) α_k

D) z