Data Analytics EEE 4774 & 6777

Module 4 - Classification

Generative Models

Spring 2022

Classification Problem & Approaches



1. Generative Models: (Likelihood, Prior, Posterior) modeling to first obtain likelihood and prior, and then posterior

$$p(C_n|\mathbf{x}_n) = \frac{p(\mathbf{x}_n|C_n)p(C_n)}{p(\mathbf{x}_n)}$$

2. Discriminative Models: (Posterior) modeling to directly discriminate classes

$$p(C_n|\mathbf{x}_n) = y(\mathbf{x}_n) = f(\mathbf{w}^T\mathbf{x}_n + w_0)$$

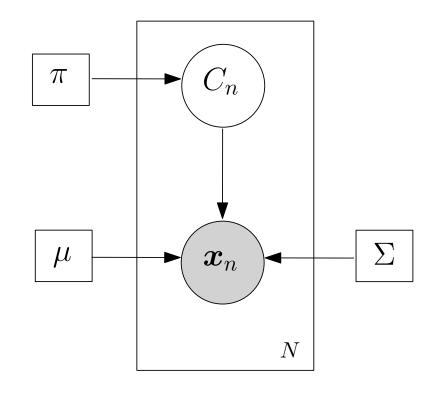
Generative Models

$$p(C_n|\mathbf{x}_n) = \frac{p(\mathbf{x}_n|C_n)p(C_n)}{p(\mathbf{x}_n)}$$

$$\max p(C_n|\mathbf{x}_n) = \max p(\mathbf{x}_n|C_n) p(C_n)$$

e.g., Gaussian mixture model, Naïve Bayes,

<u>Deep Generative Models:</u>
Variational autoencoder (VAE),
Generative adversarial network (GAN)



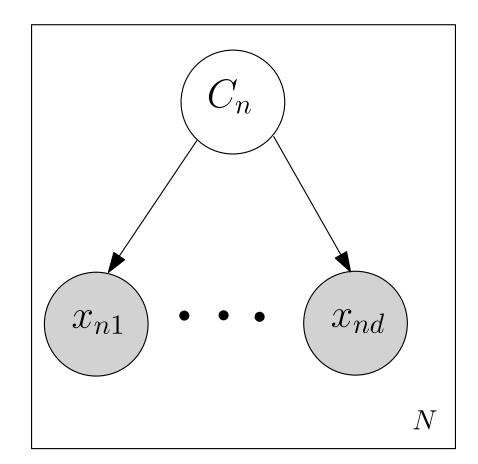
Naïve Bayes

• Assumption: Conditioned on the class C_n input variables x_{n1}, \dots, x_{nd} are independent

$$p(\mathbf{x}_n|C_n) = \prod_{i=1}^d p(\mathbf{x}_{ni}|C_n)$$

$$p(\mathbf{x}_n) \neq \prod_{i=1}^d p(x_{ni})$$

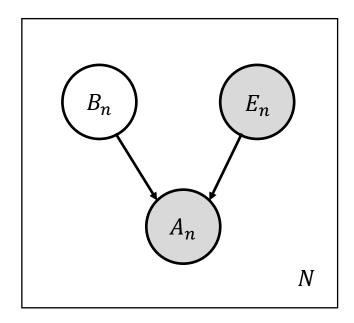
- Simplified analysis: Univariate models instead of multivariate (esp. for large *d*)
- Limitation: Assumption does not hold in general



Example: Burglary-Alarm-Earthquake

Sally's burglar Alarm is sounding.

Has she been Burgled, or was the alarm triggered by an Earthquake?



Training & Model Building

Fit a Bernoulli distribution for each alarm case (4 cases) using training data

Alarm = 1	Burglar	Earthquake
0.9999	1	1
0.99	1	0
0.99	0	1
0.0001	0	0

Prior probabilities for Burglary and Earthquake estimated from training data

$$p(B = 1) = 0.01$$
 and $p(E = 1) = 0.000001$

Testing

$$p(B=1|A=1,E=0) = \frac{p(B=1,A=1,E=0)}{\sum_{B} p(B,A=1,E=0)} \approx 0.99$$

$$p(B=1|A=1,E=1) = \frac{p(B=1,A=1,E=1)}{\sum_{B} p(B,A=1,E=1)} \approx 0.01$$

$$p(B = 1|A = 1) = p(B = 1|A = 1, E = 1)p(E = 1)$$
$$+p(B = 1|A = 1, E = 0)p(E = 0)$$

 ≈ 0.99

Earthquake explains away the fact that alarm is ringing.