



1 Which of the following is assumed when classifying using the Naive Bayes method?

- A. The features have strong correlation with each other
- B. The features are all normalized before classification
- **C RIGHT.** The features are independent/conditionally independent of each other
- D. The data is low dimensional in nature

2 The primary role of Principal component analysis is

- A. Classification
- B. Regression
- C. Clustering
- **D RIGHT.** Dimensionality Reduction

3 Which direction does PCA use to perform dimensionality reduction?

- **A RIGHT.** Direction of maximum variance
- B. Direction of minimum variance
- C. Direction of maximum mean
- D. Direction of minimum mean

4 Which of the following is/are true about PCA?

- **A RIGHT.** The principal components are orthogonal to each other
- **B RIGHT.** There can be only one principal component in PCA
- C. There are always lesser principal components than initial dimensions
- D. None of the above

Accepted Answers:

- The principal components are orthogonal to each other
 - There are always lesser principal components than initial dimensions
-

▼ 5

The given probability distribution is $P(X|\mu) = \mu^x(1 - \mu)^{(1-x)}$. Which of the following is True?

- A. $E[x] = \mu$
- B. Variance, $var(x) = \mu(1 - \mu)$
- C. (Cross) Entropy: $-\mu \ln \mu - (1 - \mu) \ln(1 - \mu)$
- **D RIGHT.** All the above

- [Bernoulli distribution](#). But here the range of x is not given in the question.

The outcome of a series of N tosses is given by the vector X which contains either 1 or 0 in each element. Assuming that the probability of a heads is μ and that each toss is independent, the probability $P(X|\mu)$ is given by

By above data answer the following questions 6 & 7

6.

What would be the best name for the distribution

- A RIGHT.

$$P(X|\mu) = \prod_{i=1}^N \mu^{x_i} (1 - \mu)^{(1-x_i)}$$

- B.

$$P(X|\mu) = \prod_{i=1}^N \mu^{x_i}$$

- C.

$$P(X|\mu) = \prod_{i=1}^N (1 - \mu)^{(1-x_i)}$$

- D.

None of the above

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If we are given the outcomes of a series of N tosses, the maximum likelihood estimation of the parameter μ

- A.

$$\mu_{ML} = \frac{1}{N} \sum_{n=1}^N x_n^2$$

- B.

$$\mu_{ML} = \frac{1}{N} \sum_{n=1}^N nx_n$$

- C.

$$\mu_{ML} = \frac{1}{N} \sum_{n=1}^N nx_n^2$$

- **D RIGHT.**

$$\mu_{ML} = \frac{1}{N} \sum_{n=1}^N x_n$$

8 Which of the following is true for the Gaussian distribution?

- **A.** The mean of the distribution is always zero
- **B RIGHT.** In higher dimensions, covariance is represented by a matrix
- **C RIGHT.** The Gaussian is always symmetric about the mean
- **D RIGHT.** The mean of the Gaussian is the same as its mode

9 During linear regression, the maximum likelihood estimate of the parameters would be

- **A.** Greater than the least squares parameters
- **B.** Lesser than the least squares parameters
- **C RIGHT.** Same as the least squares parameters
 - [In a linear model, if the errors belong to a normal distribution the least squares estimators are also the maximum likelihood estimators.](#)
- **D.** Have no connection with the least squares parameters

10 Suppose you wanted to model the probability distribution of a set of people liking a particular dish. Which form of distribution would be most apt? (Hint: They can either like the dish or they can't.)

- **A.** Normal Distribution
- **B RIGHT.** Bernoulli Distribution
- **C.** Exponential Distribution
- **D.** Log-Normal Distribution

