

AINT351 - Distributions

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More on Distributions

Expected value

- Discrete case:

$$E(X) = \sum_{\text{all } x} x_i p(x_i)$$

- The sample mean:

$$\mu = \frac{\sum_{i=1}^n x_i}{n}$$

- Continuous case:

$$E(X) = \int_{\text{all } x} x_i p(x_i) dx$$

Variance

- Calculate mean before calculating the variance

- Discrete case:

$$Var(X) = \sigma^2 = \sum_{\text{all } x} (x_i - \mu)^2 p(x_i)$$

- The sample variance:

$$\sigma^2 = \frac{\sum_{i=1}^n (x_i - \mu)^2}{n-1}$$

- Continuous case:

$$Var(X) = \sigma^2 = \int_{-\infty}^{\infty} (x_i - \mu)^2 p(x_i) dx$$

Covariance: joint prob

- The covariance measures the strength of the linear relationship between two variables
- This comes into practise when having two deminsional data that isn't independant

- The covariance is given by:

$$\sigma_{xy} = \frac{E[(x - \mu_x)(y - \mu_y)]}{N} = \sum_{i=1}^N (x_i - \mu_x)(y_i - \mu_y) P(x_i, y_i)$$

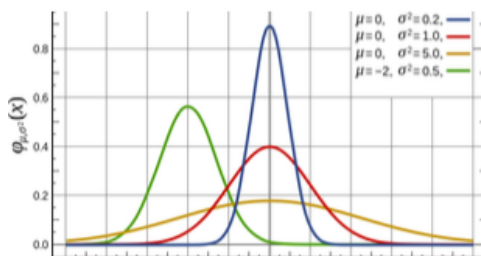
- The sample covariance:

$$\sigma_{xy} = \frac{\sum_{i=1}^n (x_i - \bar{X})(y_i - \bar{Y})}{n - 1}$$

- E = expectation

Normal distributions

- Quite often continuous values will be characterised by a Normal (or Gaussian) distribution
- Sufficient statistics:
 - mean
 - standard deviation
- if you change the standard deviation you change the spread of the curve

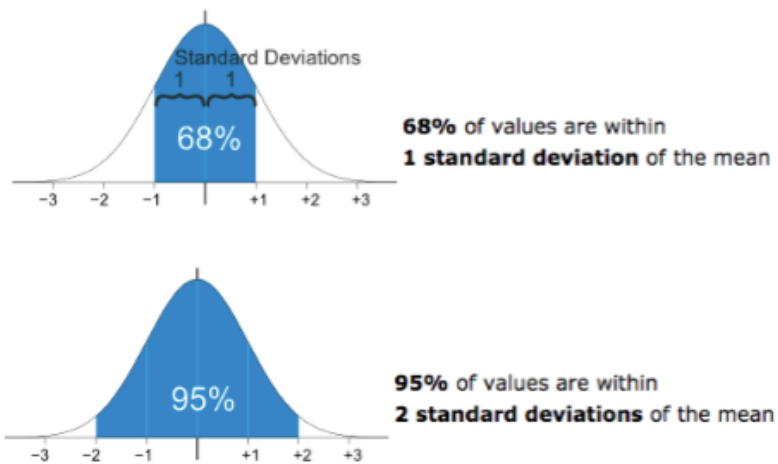


Sufficient statistics:

- mean
- standard deviation

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-(x-\mu)^2/2\sigma^2} \quad \mu, \sigma \text{ are given constants}$$

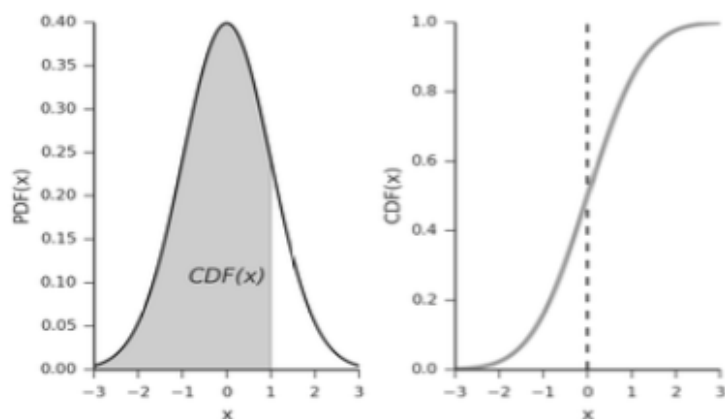
Effect of standard deviation



Cumulative distribution function

- For a continuous random variable X
 - The cumulative distribution function $CDF(x)$ represents the area under the probability density function $P(x)$ to the left of X
- not looking at the point x , what is the likelihood of all that is happening at that point and at all points previous
- area from that point - previous

$$CFD(x) = P(X \leq x)$$



Exponential distribution

- when the probability is above 0
 - not gaussian (can or is negative)

Interpreting covariance

- If we calculate the covariance between two random variables
 - Will have two sets of results for everything, mean etc
- If $\text{cov}(X, Y)$ is positive
 - Positively Correlated
- If $\text{cov}(X, Y)$ is negative
 - Negatively correlated
- if $\text{cov}(X, Y)$ is 0
 - Independent

To remember

- Mostly caring about the area under a curve

to do

- look at all functions, try to find videos explaining them

– go to the math place in library if can't understand