

Descriptive statistics

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Density & distribution functions

Cumulative distribution function (C.D.F)

(distribution function) F

$$F(x) = P(X \leq x) = P(\omega \in \Omega : X(\omega) \leq x)$$

when $X = \{x_1, x_2, \dots\}$

$$F(x) = P(X \leq x) = \sum_{x_i \leq x} P(X \leq x_i)$$

$$\lim_{x \rightarrow -\infty} F(x) = 0 \quad \text{and} \quad \lim_{x \rightarrow \infty} F(x) = 1$$

For a continuous case.

$$F(x) = \int_{-\infty}^x f(y) dy.$$

if we have several random variables.

Sample space Ω : exoplanets within 50 of
random variables could be $\{x_1, x_2, x_3\}$ x_1 : exoplanet mass x_2 : radii x_3 : surface temperature x_4 : orbital axis

$$X = (X_1, \dots, X_k)$$

Vector of random variable: (X_1, \dots, X_k)

$$F(x_1, \dots, x_k) = P(X_1 \leq x_1, \dots, X_k \leq x_k) \\ = P(\omega \in \Omega : X_1(\omega) \leq x_1, \dots, X_k(\omega) \leq x_k)$$

Marginal distribution:

For an individual X_i :

$$F(x_i) = P(X_i \leq x_i)$$

Expected Means

The Mean of a random variable is defined as a weighted average where the weight is obtained from the associated probabilities

$$E(X) = \sum x_i P(X = x_i)$$

if $Y = h(X)$: real-valued function

$$E(h(X)) = \sum h(x_i) P(X = x_i)$$

$$P(X = x_j) = P(\omega = \omega_j) = 2^{-j-1} \text{ for } j = 0, 1, \dots, n-1$$

$$\frac{1}{2} (W - K) - \frac{1}{2} (W - K) = 0$$

$$E[W] = ?$$

$E[W]$ cannot be defined.

$$W^+ = \max(0, W)$$

$$W^- = \max(0, -W)$$

$$E[W] =$$

Which is

$$\sum_i |x_i| P(X = x_i) < \infty$$

finite

Continuous Case:

$$E(X) = \int_{-\infty}^{\infty} y f(y) dy \quad \text{provided}$$

$$E[h(X)] = \int_{-\infty}^{\infty} h(y) f(y) dy \quad \approx \int_{-\infty}^{\infty} h$$

Variance:

Quantifies the "Spread" is the S^2 centered on the Mean.

$$\sigma^2 = \text{Var}(X) = E[(X - \mu)^2]$$

$$\mu = E[X]$$

If we have X_1, X_2, \dots, X_n .

$$E\left[\sum_{i=1}^n X_i\right] = \sum_{i=1}^n E[X_i]$$

$$\text{Var}\left(\sum X_i\right) = \sum \text{Var}(X_i) + \sum_{i=1}^n \sum_{\substack{j=1 \\ i \neq j}}^n \text{Cov}(X_i, X_j)$$

$$\text{Cov}(X, Y) = E[(X - E(X))(Y - E(Y))]$$

Covariance b/w X & Y .

Measures the relationship b/w the in two random variables.

If all the X_i variables have the same $\text{Var}(X_i)$, the situation is called "homoscedastic".
the variance of the sample mean $\bar{X} = \frac{\sum X_i}{n}$

$$\text{Var}(\bar{x}) = \frac{1}{n^2} \sum \text{Var}(x_i) = \sigma^2/n.$$

Standard deviation = $\sigma = \sqrt{V}$

Standardized form:

$$X_{\text{std}} = \frac{X - \mu}{\sigma}$$



- Normalisation. — Zero mean and $\sigma = 1$
- logarithmic transformation.

Skewness

$$E[(X - E[X])^3] \rightarrow \mu_3$$

Quantile function

$F(x) \rightarrow$ estimates the value of the population distribution function.

value of x .

Astronomy: What value of x corresponds to a specified value

What fraction of galaxies have above L^* ?

↓ threshold value.

$$Q(u) = F^{-1}(u) = \inf\{y : F(y) \geq u\}$$

where $0 < u < 1$. \inf : infimum: value of y with the property of the brackets.

quantiles such as 5%, 25%, 50%, 75%, 95%

Quantile-Quantile Plots (Q-Q Plots)

Next lecture:

- * Estimators
 - * Sample Mean, Sample
 - * Variance / mean of \bar{u}
 - * Real Data;
 - * Outliers.
 - * Distribution functions
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