# CBD2204: week 4

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#### data frames & random variables

consider a data frame, consisting of rows and labeled columns:

index	random variable 1	random variable 2	random variable 3	random variable 4	
sample 1					
sample 2					
sample 3					
:	:		:	:	,

Big Data just means that we have many samples (rows)

# example (data frame)

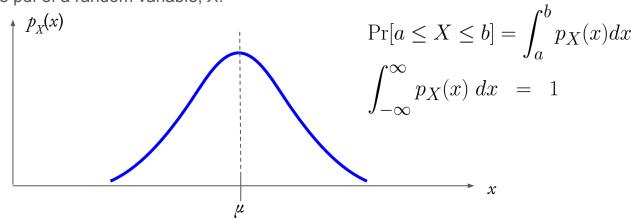
#### customer data

customer (sample index)	number of orders	total sales	gender
0001	5	238.77	М
0002	3	36.49	F
0003	8	313.28	U
0004	2	15.12	М
0005	9	1043.86	М
0006	4	422.27	F
0007	3	163.44	F

columns ⇔ random variables

#### random variables

- a random variable is a mathematical concept; it refers to a sampled quantity, whose value (when sampled)
  cannot be predicted, but behaves according to a probability distribution
- think of the probability distribution as you would a histogram, since both will have the same shape, after an infinite number of samples
- for example, for any given customer (a sample), we don't know how much that customer will spend (sales) or how many times that customer will make a purchase (number of orders), but we can characterize the sales and number of orders as random variables, i.e., *statistically*.
- **example**: here is the pdf of a random variable, *X*:



<sup>\*</sup>we assume that x (i.e., the value of the associated random variable), is a real quantity

## random variables (expectation)

• we define the **expected value** of a random variable X in terms of its pdf as:

$$E[X] = \int_{\forall x} x \, p_X(x) \, dx$$

- which captures the intuition of a "weighted average" or mean value of the random variable, it is often represented by the Greek letter  $\mu$ , and also referred to as the *first moment* of the pdf.
- the deviation of the pdf from the mean is described by the **variance**, defined as:

$$Var[X] = E[(X - \mu)^2]$$

 and gives a measure of the "spread" in values about the mean; it also called the second moment; the square-root of the variance is called the standard deviation:

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

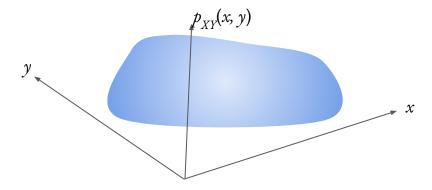
• as a consequence, the variance is sometimes represented by  $\sigma^2$ 

# random variables (cont'd)

- dependence: although they are not predictable, two random variables can have probability distributions which depend on each other
- two random variables, X and Y, are independent iff

$$p_{XY}(x,y) = p_X(x) \ p_Y(y), \ \forall \ x, \ y$$

• in which  $p_X()$  is pdf of X,  $p_Y()$  is the pdf of Y, and  $p_{XY}()$  is the joint probability distribution



#### random variables (cont'd)

- if two random variables are *independent*, then they are *uncorrelated*. However, *correlated* variables are not necessarily dependent, in the probabilistic/mathematical sense.
- we define the covariance of random variables X and Y as:

$$cov(X,Y) = E[(X-\mu_X)(Y-\mu_Y)]$$

 we can use covariance to determine the *correlation coefficient*, between two random variables:

$$\rho_{XY} = \frac{\text{cov}(X, Y)}{\sigma_X \, \sigma_Y}$$

 which gives a measure of the quality of a least-squares fit between the variables

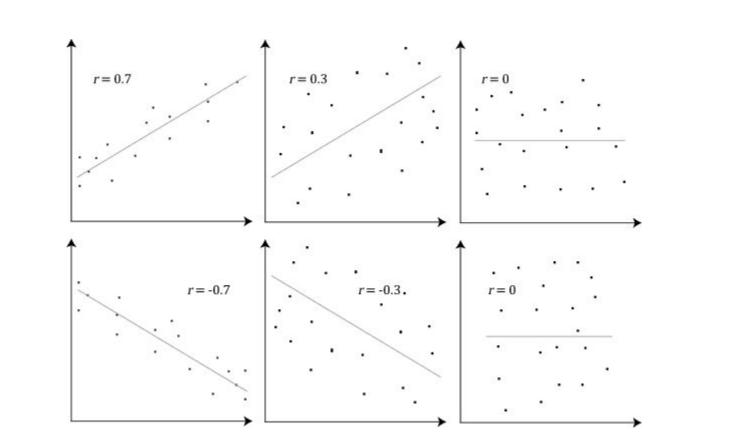
## sample calculations

$$E[X] \approx \frac{1}{N} \sum_{i=1}^{N} x_i =: m_X$$

sample version of correlation coefficient:

$$r = \frac{1}{N} \frac{\sum_{i=1}^{N} (x_i - m_X)(y_i - m_Y)}{s_X s_Y}$$

it helps if N is large!



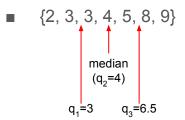
# descriptive statistics in R

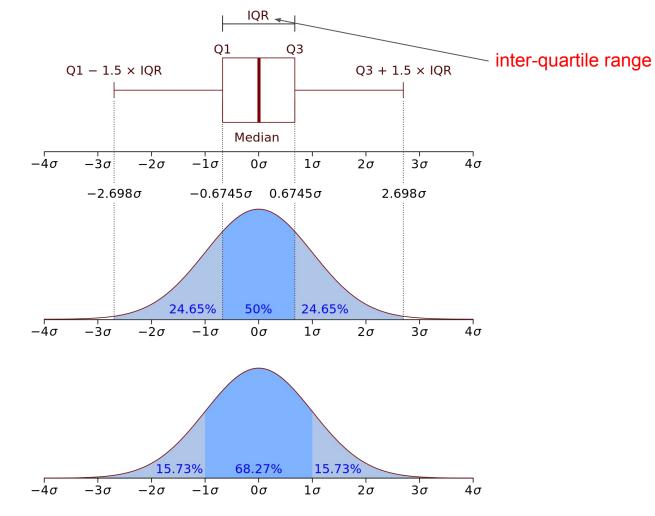
recall that the *summary*() function in *R* provides several descriptive stats about the variables (columns) in a data frame:

- mean
- median
- quartiles

for median and quartiles, data is *ordered*; for an odd-number of samples, the median is the middle sample point, and for an even number of samples, the median is the average of the smallest and largest samples.

- example: consider our *number of sales* data above: {5, 3, 8, 2, 9, 4, 3}
- ordering yields:





#### back to R

check out the following functions in *R*:

- summary()
- mean()
- median()
- cov()
- cor()