

# A statistic

(STATS201.stats 201 20: Probability and probability distributions)

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## Statistic

A **statistic** is a number that is calculated by some method from our data.

- We look at our measurements as random variables because the outcome can change each time the experiment is repeated.
- Statistics are calculated from our measurements.
- If the outcomes change, the statistics can also change!
- That means that statistics are in fact random variables!

# Sampling distribution

## Sampling distribution

Every statistic is a random variable and has therefore some probability distribution. That distribution is called the **sampling distribution** of the statistic.

The sampling distribution depends on

- The **probability distribution of the measurements** that the statistic is calculated for.
- The **number of measurements**.

When certain criteria are fulfilled the sampling distribution of some statistics follow certain known types. Statistical inference normally relies on that fact.

## Example

Let  $X_1$  and  $X_2$  be random variables that describe the outcome when a dice is thrown.

Below are shown all possible outcomes of the statistic  $X_1 + X_2$ .

					(6,1)						
			(4,1)	(5,1)	(5,2)	(6,2)					
		(3,1)	(3,2)	(4,2)	(4,3)	(5,3)	(6,3)				
	(2,1)	(2,2)	(2,3)	(3,3)	(3,4)	(4,4)	(5,4)	(6,4)			
(1,1)	(1,2)	(1,3)	(1,4)	(2,4)	(2,5)	(3,5)	(4,5)	(5,5)	(6,5)		
				(1,5)	(1,6)	(2,6)	(3,6)	(4,6)	(5,6)	(6,6)	
2	3	4	5	6	7	8	9	10	11	12	

# Example

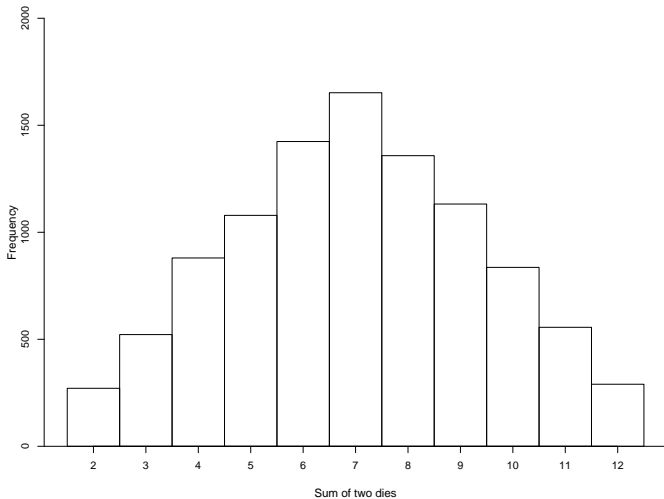


Figure: Simulation of 10000 throws of two dice.

# Expected value of the sum of two random variables

Formulas for the expected value of random variables

If  $X$  and  $Y$  are two random variables, then

$$E[X + Y] = E[X] + E[Y]$$

$$E[X - Y] = E[X] - E[Y]$$

# Variance of the sum of two random variables

## Formulas for the variance of random variables

If  $X$  and  $Y$  are two independent random variables then

$$\text{Var}[X + Y] = \text{Var}[X] + \text{Var}[Y]$$

$$\text{Var}[X - Y] = \text{Var}[X] + \text{Var}[Y]$$

# Expected value and variance of the mean

## Expected value and variance of the mean

If  $X_1, \dots, X_n$  are independent and identically distributed random variables with expected value  $E[X_i] = \mu$  and variance  $\text{Var}[X_i] = \sigma^2$ , then the following holds for the mean of them, denoted  $\bar{X}$ :

$$E[\bar{X}] = \mu$$
$$\text{Var}[\bar{X}] = \frac{\sigma^2}{n}$$



# Standard error

## Standard error

If  $\bar{X}$  is the mean of  $X_1, \dots, X_n$ , independent and identically distributed random variables with variance  $\text{Var}[X_i] = \sigma^2$ , then their **standard error** is

$$\sigma/\sqrt{n}$$

It is the standard deviation of the mean of the measurements.

# The probability distribution of the mean of normally distributed random variables

The probability distribution of the mean of normally distributed random variables

If  $X_1, \dots, X_n$  are normally distributed random variables with mean  $\mu$  and variance  $\sigma^2$ , then  $\bar{X}$  follows also a normal distribution, with mean  $\mu$  and variance  $\sigma^2/n$ .

That is if  $X_i \sim N(\mu, \sigma^2)$  then  $\bar{X} \sim N(\mu, \sigma^2/n)$ .

# Central limit theorem

## Central limit theorem

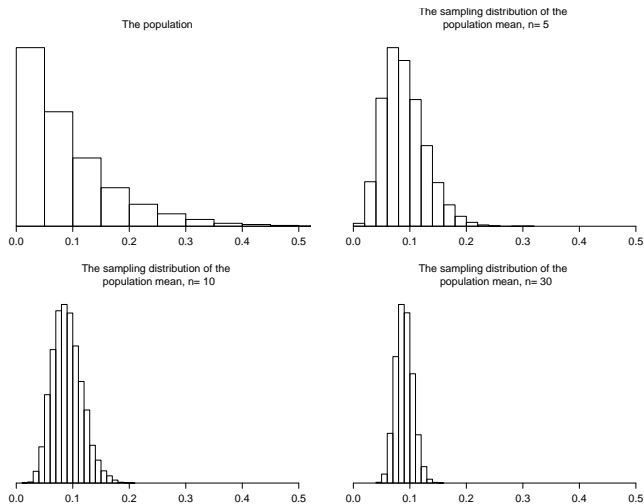
If  $X_1, \dots, X_n$  are independent and identically distributed variables then  $\bar{X}$  follows normal distribution with mean  $\mu$  and variance  $\sigma^2/n$

$$\bar{X} \sim N(\mu, \sigma^2/n)$$

if  $n$  is large enough.

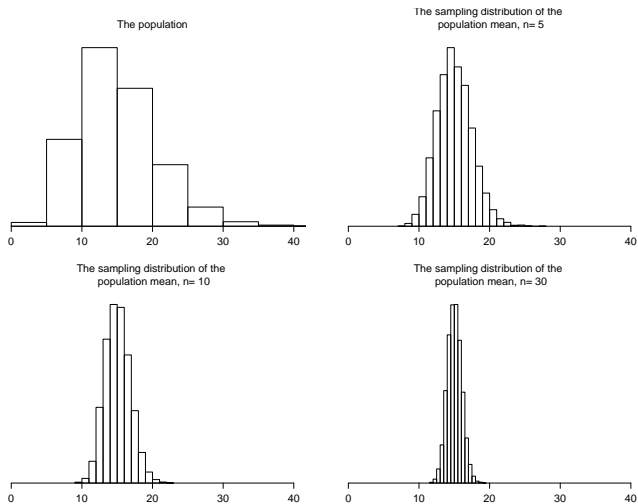
Notice that we do not need to know the probability distribution of the measurements!

# Central limit theorem



**Figure:** The sampling distribution of a mean when the random variables follow a very skewed distribution.

# Central limit theorem



**Figure:** The sampling distribution of a mean when the random variables follow a slightly skewed distribution.

# Estimators and test statistics

There are two groups of important statistics.

- **Estimators** estimate the parameters of the probability distribution that the random variables follow.  
Example: Estimator that estimates  $\mu$  when the measurements are normally distributed.  
Example: Estimator that estimates  $p$  when the measurements are binomially distributed.
- **Test statistics** allow us to make statistical inference.  
Example: Test statistic that allows us to infer whether the variance of two populations is the same.  
Example: Test statistic that allows us to infer whether the mean of a population differs from 20.