# Beamer presentation using R Markdown

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### Section 1

# Central Limit Theorem (CLT)

## Central Limit Theorem (CLT)

Let  $X_1, X_2, \ldots, X_n$ , be a sequence of independent and identically distributed (iid) random variables (rv) from a distribution with mean  $\mu$  and finite variance  $\sigma^2$ .

For large n,

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i \sim \mathcal{N}(\mu, \sigma^2/n)$$

where  $\dot{\sim}$  indicates convergence in distribution. Equivalently

$$S_n = \sum_{i=1}^n X_i \sim \mathcal{N}(n\mu, n\sigma^2)$$

The CLT supports the normal approximation to the distribution of a rv that can be viewed as the sum of other rv.

### Approximation with CLT: application

The approximation above is useful in statistics for computing some quantities. For instance, let X and Y be two independent Binomial rv, such that  $X \sim \operatorname{Bin}(n,p)$  and  $Y \sim \operatorname{Bin}(m,q)$ .

If we are interested in computing the probability P(X > Y) the Normal approximation is the simplest way to compute it. We can approximate:

$$X \approx \mathcal{N}(np, np(1-p)), \qquad Y \approx \mathcal{N}(mq, mq(1-q)).$$

Then, by using a well known probability result, the difference W=X-Y of two independent normal distributions with means  $\mu_X,\mu_Y$  and variances  $\sigma_X^2,\sigma_Y^2$ , respectively, is **still a normal distribution** with mean  $\mu_W=\mu_X-\mu_Y$  and variance  $\sigma_W^2=\sigma_X^2+\sigma_Y^2$ .

In such a case,

$$\mu_W = \mu_X - \mu_Y = np - mq,$$
  $\sigma_W^2 = \sigma_X^2 + \sigma_Y^2 = np(1-p) + mq(1-q).$ 

#### Section 2

CLT application to waterpolo goals

Tomorrow two professional Italian waterpolo teams, Posillipo and Pro Recco, compete against each other. Let X and Y be the random goals scored by Posillipo and Pro Recco, respectively.

We assume that X, Y follow two independent Binomial distributions. Thus, X and Y represent the number of shots converted in goal on the total number of shots n, m made by Posillipo and Pro Recco, with probabilities p and q, respectively.

Before the match, the number of shots is *unknown*. In what follows, we adopt a simplification, and we treat the quantities p, q, m, n as *known*, for instance fixing them upon historical experience:

$$p = 0.5, q = 0.7, n = 20, m = 20.$$

We want to investigate the Posillipo probability of winning the next match against Pro Recco, that is

$$P(X > Y) = P(X - Y > 0) = ?$$

So, let W be the r.v., such that W=X-Y. We could compute the law of this rv but using the Normal approximation,

 $W \approx \mathcal{N}(\mu_W = \mu_X - \mu_Y, \sigma_W^2 = \sigma_X^2 + \sigma_Y^2)$ , we can esily compute such a probability of interest.

```
p <- 0.5
q <- 0.7
n <- m <- 20
mW <- p * n - q * m
sdW <- sqrt( n * p * (1 - p) + m * q * (1 - q) )
# Probability that W = X-Y > O (Posillipo win the match)
PWin_P <- pnorm(0, mean = mW, sd = sdW, lower.tail = FALSE)
PWin_P</pre>
```

## [1] 0.09362452

Here, we show the probability mass functions of X and Y and the probability density function of W=X-Y.

```
# pdf of W
curve(dnorm(x, mW, sdW),
      xlim = c(-12, 20), ylim = c(0, 0.3),
      xlab = "", ylab = "", cex.lab = 1.25)
# pmf of X and Y
points(0 : 20, dbinom(0 : n, n, p), pch = 21, bg = 1, col = "blue")
points(0 : 20, dbinom(0 : m, m, q), pch = 21, bg = 2)
segments(0, -0.01,
                           \#(x \ 0, \ y \ 0)
         0, dnorm(0, mW, sdW), \#(x 1, y 1)
         1wd = 2
text(14.25, 0.22, "Y", cex = 2, col = 2)
text(10, 0.2, "X", cex = 2, col = "blue")
text(-4, 0.15, "X - Y", cex = 2, col = 1)
```

