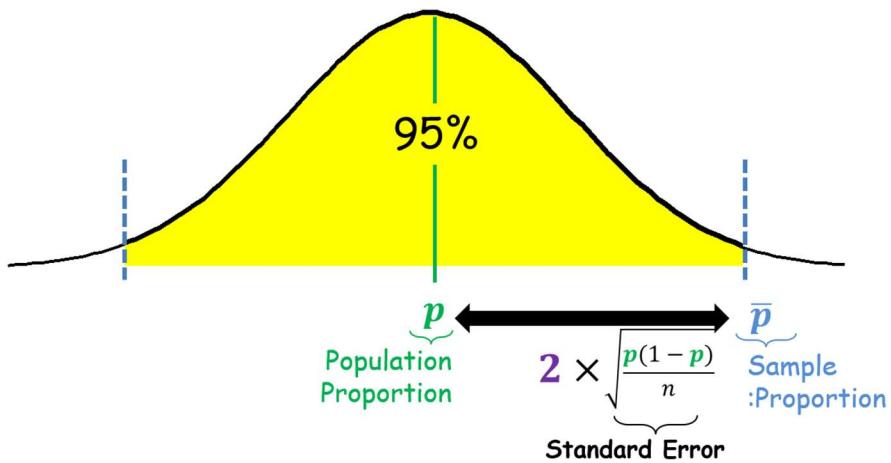


Deriving the Confidence Interval

Recall, the **sample proportion (\bar{p})** follows a Normal Distribution.

<ul style="list-style-type: none"> - The mean is given by the population proportion (p) - The standard deviation is given by $SD[\bar{p}] = \sqrt{\frac{p(1-p)}{n}}$ <p>(This term is also called Standard Error)</p>	
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According to the 95% rule for a Normal Distribution, **95% of data (e.g. sample proportions) fall within 2 standard deviations (or standard errors) of the population proportion.**



If we assume our sample proportion falls in this 95% range,

then the difference between **the sample proportion (\bar{p})** and its **population proportion (p)** is given by at most **2 × Standard Error**.

$$|p - \bar{p}| \leq 2 \times \sqrt{\frac{p(1-p)}{n}}$$

Standard Error

Then we can rewrite the expression in terms of p . We have the following expression.

$$\rightarrow \bar{p} - 2 \times \sqrt{\frac{p(1-p)}{n}} \leq p \leq \bar{p} + 2 \times \sqrt{\frac{p(1-p)}{n}}$$

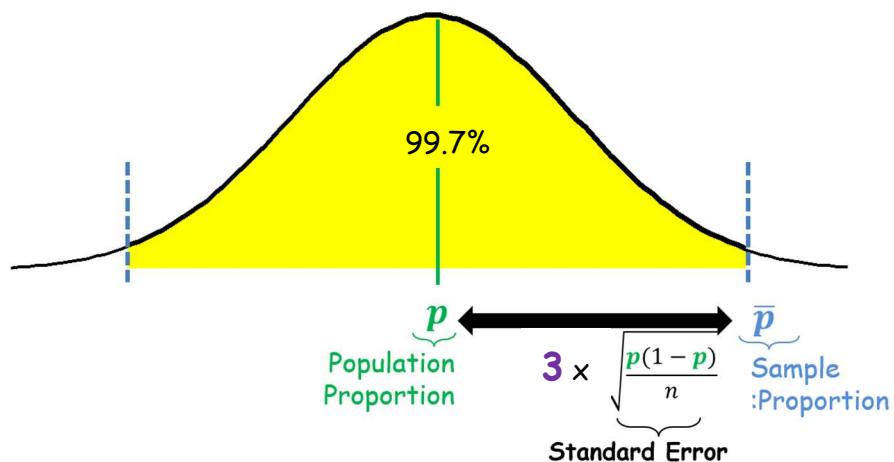
Standard Error

The expression provides an **interval** that captures the **population proportion** with a 95% probability.

Deriving the Confidence Interval

If you want to increase the probability that the interval captures the population proportions, we can use the 99.7% rule for Normal Distribution.

According to the 99.7% rule for a Normal Distribution, **99.7% of data (e.g. sample proportion) fall within 3 standard deviations (or standard errors) of the population proportion.**



If we assume our sample proportion falls in this 99.7% range,

then the difference between **the sample proportion (\bar{p})** and its **population proportion (p)** is given by at most **$3 \times \text{Standard Error}$** .

$$|p - \bar{p}| \leq 3 \times \sqrt{\frac{p(1-p)}{n}}$$

Standard Error

Then we can rewrite the expression in terms of p . We have the following expression.

$$\rightarrow \bar{p} - 3 \times \sqrt{\frac{p(1-p)}{n}} \leq p \leq \bar{p} + 3 \times \sqrt{\frac{p(1-p)}{n}}$$

Standard Error

The expression provides an **interval** that captures the **population proportion** with a 99.7% probability.

Deriving the Confidence Interval

In general, if you want to construct an interval that captures the **population proportions** with a desired probability, you only need to multiply with the standard error with the right **constant**.

$$\bar{p} - \text{constant} \times \underbrace{\sqrt{\frac{p(1-p)}{n}}}_{\text{Standard Error}} \leq p \leq \bar{p} + \text{constant} \times \underbrace{\sqrt{\frac{p(1-p)}{n}}}_{\text{Standard Errpr}}$$

The value of this constant depends on the desired probability level and is obtained from the **Normal distribution**. For this reason, it is called the **z-critical value**.

$$\bar{p} - \frac{z_c}{\text{critical value}} \times \underbrace{\sqrt{\frac{p(1-p)}{n}}}_{\text{Standard Error}} \leq p \leq \bar{p} + \frac{z_c}{\text{critical value}} \times \underbrace{\sqrt{\frac{p(1-p)}{n}}}_{\text{Standard Error}}$$

Therefore, the expression $\bar{p} \pm \frac{z_c}{\text{critical value}} \times \underbrace{\sqrt{\frac{p(1-p)}{n}}}_{\text{Stamdard Error}}$ provides an interval estimate

that gives a range of plausible values for the **population proportion**.

Since the probability level used to determine the **critical value** is called the **confidence level**, this interval estimate is formally known as a **confidence interval**.

One more thing: Since the standard error depends on the **population proportion (p)** which is **unknown**, we need to replace **p** by **the sample proportion (\bar{p})**.

$$\sqrt{\frac{p(1-p)}{n}} \rightarrow \sqrt{\frac{\bar{p}(1-\bar{p})}{n}}$$

Then the **confidence interval** is written as:

$$\bar{p} \pm \frac{z_c}{\text{critical value}} \times \underbrace{\sqrt{\frac{\bar{p}(1-\bar{p})}{n}}}_{\text{Standard Error}}$$