

1. A scientist wishes to test if a new treatment has a better cure rate than the traditional treatment which cures only 60% of the patients. In order to test whether the new treatment is more effective or not, a test group of 20 patients were given the new treatment. Assume that each personal result is independent of the others.

CDF @ x

n	x	p
20	0	0.000
	1	0.000
	2	0.000
	3	0.000
	4	0.000
	5	0.002
	6	0.006
	7	0.021
	8	0.057
	9	0.128
	10	0.245
	11	0.404
	12	0.584
	13	0.750
	14	0.874
	15	0.949
	16	0.984
	17	0.996
	18	0.999
	19	1.000

Trying to decide: cure rate $p \leq 0.60$ vs. $p > 0.60$.

- a) If the new treatment has the same success rate as the traditional, what is the probability that at least 14 out of 20 patients (14 or more) will be cured?

- b) Suppose that 14 out of 20 patients in the test group were cured. Based on the answer for part (a), is there a reason to believe that the new treatment has a better cure rate than the traditional treatment?

- c) If the new treatment has the same success rate as the traditional, what is the probability that at least 17 out of 20 patients (17 or more) will be cured?

- d) Suppose that 17 out of 20 patients in the test group were cured. Based on the answer for part (c), is there a reason to believe that the new treatment has a better cure rate than the traditional treatment?

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A **null hypothesis**, denoted by H_0 , is an assertion about one or more population parameters. This is the assertion we hold as true until we have sufficient statistical evidence to conclude otherwise.

The **alternative hypothesis**, denoted by H_1 , is the assertion of all situations *not* covered by the null hypothesis.

The test is designed to assess the strength of the evidence against the null hypothesis.

	H_0 true	H_0 false
Accept H_0 (Do NOT Reject H_0)	☺	Type II Error
Reject H_0	Type I Error	☺

$$\alpha = \text{significance level} = P(\text{Type I Error}) = P(\text{Reject } H_0 \mid H_0 \text{ is true})$$

$$\beta = P(\text{Type II Error}) = P(\text{Do Not Reject } H_0 \mid H_0 \text{ is NOT true})$$

$$\text{Power} = 1 - P(\text{Type II Error}) = P(\text{Reject } H_0 \mid H_0 \text{ is NOT true})$$

Testing Hypotheses about a Population Proportion p

Null	vs.	Alternative	
$H_0 : p \geq p_0$		$H_1 : p < p_0$	Left – tailed.
$H_0 : p \leq p_0$		$H_1 : p > p_0$	Right – tailed.
$H_0 : p = p_0$		$H_1 : p \neq p_0$	Two – tailed.

Test Statistic:
$$z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0 \cdot (1 - p_0)}{n}}}$$

$$z = \frac{Y - n \cdot p_0}{\sqrt{n \cdot p_0 \cdot (1 - p_0)}}$$

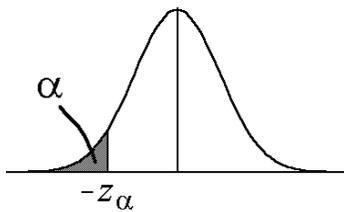
where Y is the number of S's in n independent trials.

Rejection Region:

$$H_0 : p \geq p_0$$

$$H_1 : p < p_0$$

Left - tailed.



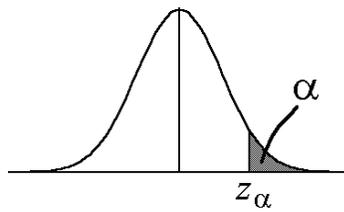
Reject H_0 if

$$Z < -z_\alpha$$

$$H_0 : p \leq p_0$$

$$H_1 : p > p_0$$

Right - tailed.



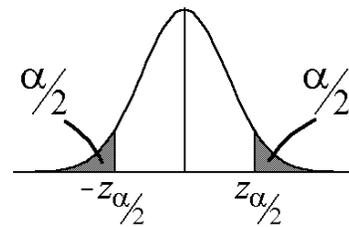
Reject H_0 if

$$Z > z_\alpha$$

$$H_0 : p = p_0$$

$$H_1 : p \neq p_0$$

Two - tailed.



Reject H_0 if

$$Z < -z_{\alpha/2}$$

or

$$Z > z_{\alpha/2}$$

If the value of the Test Statistic falls into the Rejection Region,

then Reject H_0

otherwise, Accept H_0 (Do NOT Reject H_0)

The **P-value (observed level of significance)** is the probability, computed assuming that H_0 is true, of obtaining a value of the test statistic as extreme as, or more extreme than, the observed value.

(The smaller the p-value is, the stronger is evidence against H_0 provided by the data.)

P-value $> \alpha$ Do NOT Reject H_0 (Accept H_0).

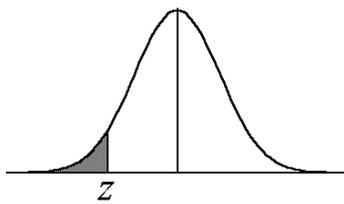
P-value $< \alpha$ Reject H_0 .

Computing P-value:

$$H_0 : p \geq p_0$$

$$H_1 : p < p_0$$

Left - tailed.

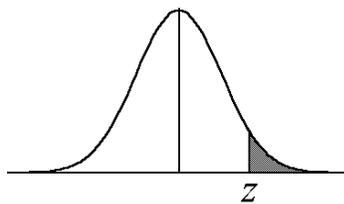


Area to the left of the observed test statistic

$$H_0 : p \leq p_0$$

$$H_1 : p > p_0$$

Right - tailed.

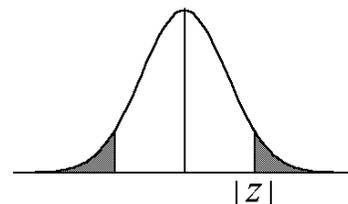


Area to the right of the observed test statistic

$$H_0 : p = p_0$$

$$H_1 : p \neq p_0$$

Two - tailed.



$2 \times$ area of the tail

c) Find the p-value of the appropriate test.

d) Using the p-value from part (c), state your decision (Accept H_0 or Reject H_0) at $\alpha = 0.08$.

