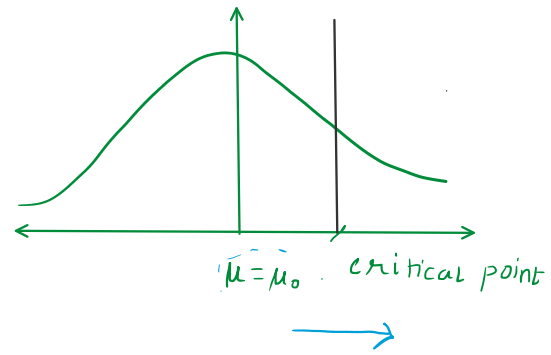




To test:  $H_0: \mu = \mu_0$  vs  $H_1: \mu > \mu_0$ .  
 To test:  $H_0: \mu = \mu_0$  vs  $H_1: \mu < \mu_0$ .  
 ↳ one-tailed test.



To test:  $H_0: \mu = \mu_0$  vs  $H_1: \mu \neq \mu_0$ .  
 ↳ Two-tailed test.

Note: In testing of Hypothesis we can have 2 types of error situations:-

		<u>actual scenario</u>	
		$H_0$ is true	$H_0$ is false
depends on the test outcome	Accept $H_0$ .	☑	Type II Error
	Reject $H_0$ .	Type I Error	☑

Type I Error: Reject  $H_0$  when  $H_0$  is true.

Type II Error: Accept  $H_0$  when  $H_0$  is false.

In practice since the true scenario is unknown, it is not possible to know if either of these errors have been committed. Hence, since minimizing these errors is not possible, we will try to minimize the probability of occurrence of these errors.

Suppose:  $\Omega$ : sample space (all possible values of  $\bar{x}$ ).

$W$ : Rejection Region / Critical region.

$\Omega - W = W^c$ : Acceptance Region.

Type I Error:  $\alpha = P[\bar{x} \in W | H_0]$

Type II Error:  $\beta = P[\bar{x} \in W^c | H_1]$

Type II Error:  $\beta = P[\bar{X} \in W^c | H_1]$

$1 - \beta = P[\bar{X} \in W | H_1]$   $\xrightarrow{\text{true scenario}}$  = Power of test.

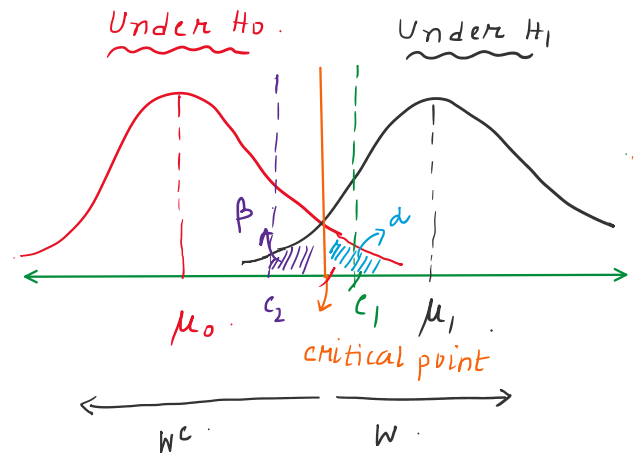
$\hookrightarrow H_0$  will be rejected based on the test.

To test:  $H_0: \mu = \mu_0$  vs  $H_1: \mu = \mu_1$  ( $\mu_1 > \mu_0$ ) [ $\sigma^2$  is known]

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

Under  $H_0: \bar{X} \sim N(\mu_0, \frac{\sigma^2}{n})$

Under  $H_1: \bar{X} \sim N(\mu_1, \frac{\sigma^2}{n})$



Note: The choice of critical point determines the values of Type I Error and Type II Error.

No choice of critical point is possible that minimizes both these errors.

[If we try to choose a critical point to minimize one of the errors, the other will increase]