

2.4.1 \rightarrow Neyman-Pearson theory

\rightarrow most powerful tests

\rightarrow limit super to simple hypotheses for H_0 & H_1

\rightarrow UMP tests have highest power (relative to all) other α -level tests

\rightarrow Neyman-Pearson Theorem

\rightarrow A test of size α is a most powerful (i.e. best critical region) iff

	<u>Condition</u>	<u>Critereon</u>	
when sample \in critical region	$\frac{L(\theta_1)}{L(\theta_0)} \leq k$	$\frac{L(\theta_1)}{L(\theta_0)} \leq k$	$\Rightarrow \text{PZR } \times$
----- $\&$ -----	$\frac{L(\theta_1)}{L(\theta_0)} \geq k$	$\frac{L(\theta_1)}{L(\theta_0)} \geq k$	$\Rightarrow \text{R } \checkmark$

} small ratio values indicate data favours H_0

\rightarrow Start w/ ratio of likelihoods & manipulate until a meaningful TS emerges. This helps to determine the best critical region

$$\frac{L(\theta_1)}{L(\theta_0)} \leq k$$

2.4.2 \rightarrow Uniformly most Powerful tests (UMP)

\rightarrow Overview \rightarrow usual scenario is when testing a simple hypothesis vs a composite hypothesis

\rightarrow In this case, an UMP test has the same best critical region for all possibilities described in H_1 .

\rightarrow Essentially, apply the Neyman-Pearson theory & analyze critical region to see if it changes based on different possible values under H_1

\rightarrow To get the final critical region, we $\rho(\text{Reject} | H_1) = \alpha$
to find c in $P(T.S. \leq c | H_0) = \alpha$
 \hookrightarrow use distribution knowledge