

HYPOTHESIS TESTING.

ref $\begin{cases} 1. \text{Signmund Brandt} \\ 2. \text{Glen Cowan} \end{cases}$

We have seen from the disease test example (application of Bayes theorem) that given a test and two hypotheses two types of errors are possible

Type-I α false positive (error rate α)

Type-II β false negative. (error rate β)

In the case of a signal-background analysis the two hypothesis are

$H_0 \rightarrow$ signal not present : Null hypothesis

$H_1 \rightarrow$ signal present : Alternative hypothesis
(can be simple or composite)

Our goal is to devise tests that minimize both the error rates.

Not possible simultaneously \rightarrow If I want to reject more background events I will reduce type I rate but I will also reject more signal events, i.e. increase the type-II error rate.

\Rightarrow For a given α design a test that minimizes β

$1 - \beta$: power of the test

designing a test means : choosing a test statistic t and defining a critical region S_c .

- If t falls in the S_c we will reject H_0 .

Examples: Covid test : $t = CT$ value
 critical region $S_c \Rightarrow t \leq 35$
 H_0 : Covid not present.

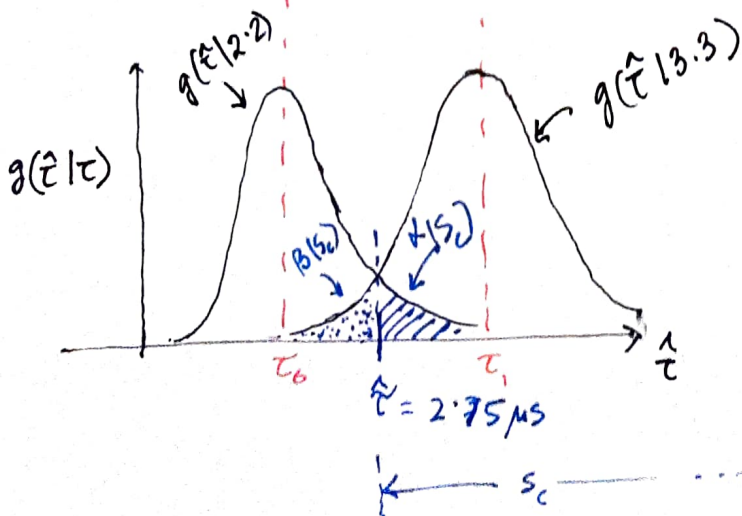
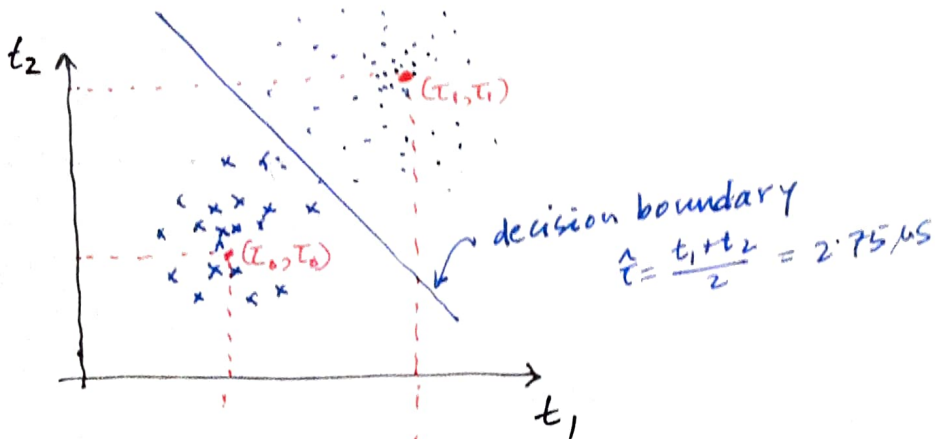
Another example :

$H_0 = \mu$ lifetime τ_0 is $2.2 \mu S$.

$H_1 = \tau_1$ is $3.3 \mu S$.

$$t \equiv \hat{t} = \frac{1}{n} \sum t_i$$

E space : $\vec{x} = (t_1, t_2, \dots, t_n)$



if $\hat{t} \geq 2.2 \mu S$

H_0 is rejected.

Note: if we try to reduce α , β increases

We want to find t s.t. $g(\hat{t} | \tau_0)$ and $g(\hat{t} | \tau_1)$ are as far apart as possible.

** Neural networks are often used to find most optimal decision boundaries $\rightarrow t = NN$ output.

We see that each hypothesis is defined by a distribution.

$$H_0: \prod_i \frac{1}{\tau_0} e^{-t_i/\tau_0}$$

$$H_1 = \prod_i \frac{1}{\tau_1} e^{-t_i/\tau_1}$$

$$\textcircled{G} \quad H_0: \frac{1}{\sqrt{2\pi}\sigma_0^2} e^{-(\hat{t} - \tau_0)^2/2\sigma_0^2}$$

$$H_1 = \frac{1}{\sqrt{2\pi}\sigma_1^2} e^{-(\hat{t} - \tau_1)^2/2\sigma_1^2}$$

$$g(\hat{t} | \tau_0)$$

$$g(\hat{t} | \tau_1)$$

$$\int_{S_c} g(\hat{t} | \tau_0) d\hat{t} = \alpha$$

$$\int_{S_c} g(\hat{t} | \tau_1) d\hat{t} = \beta$$

UMP: Uniformly most powerful test is one which maximizes $(1-\beta)$ for all possible H_1 .

* UMP may or may not exist