

MLE, LRT, Hotelling's T^2

&

Testing Hypothesis

H. Park

HUFS

순서

- 확률함수와 우도함수
- 우도함수와 최우추정량(MLE)
- 우도함수와 우도비(LRT) 검정통계량
- 다변량정규분포의 우도함수
- 다변량 정규분포의 우도비(LRT) 검정통계량
- 다변량 정규분포와 Hotelling's T^2 검정통계량
- 일반선형가정과 Hotelling's T^2 검정통계량

확률함수 $P(x|\theta)$ vs. 우도함수

Probability ft.(확률함수): $P(x|\theta)$

(e.g) 동전 2개를 던져서 앞면이 x 개 나올 확률

$$P(x | \theta) =$$

$$\theta = \frac{1}{2} \rightarrow P(x|\theta = 1/2) =$$

$$x = 2$$

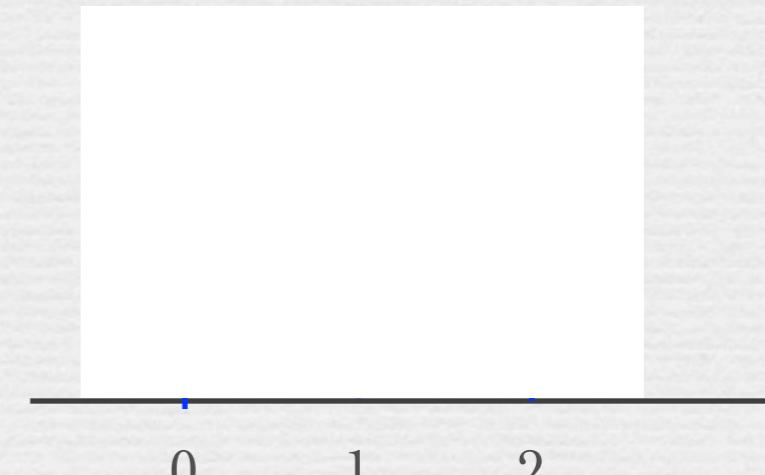
$$P(x = 2|1/2) =$$

$$x = 1$$

$$P(x = 1|1/2) =$$

$$x = 0$$

$$P(x = 0|1/2) =$$



$$P(x | \theta) = \binom{2}{x} \theta^x (1 - \theta)^{2-x}$$

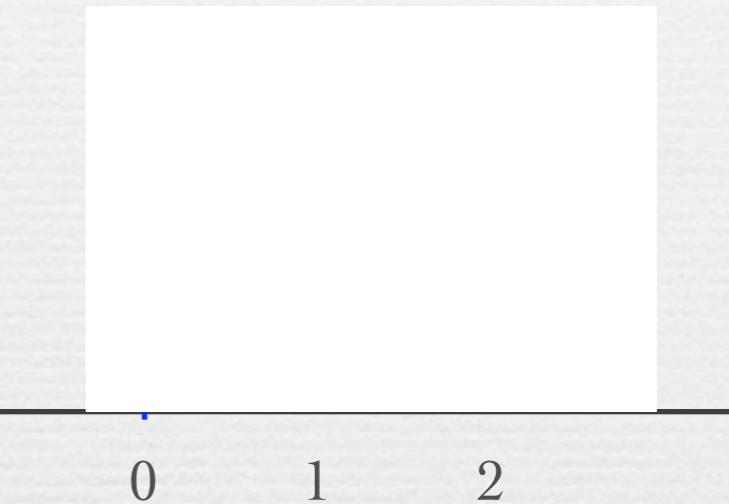
$$\theta = \frac{1}{4}$$

$$P(x|\theta = 1/2) =$$

$$x = 2 \quad P(x = 2|1/4) =$$

$$x = 1 \quad P(x = 1|1/4) =$$

$$x = 0 \quad P(x = 0|1/4) =$$



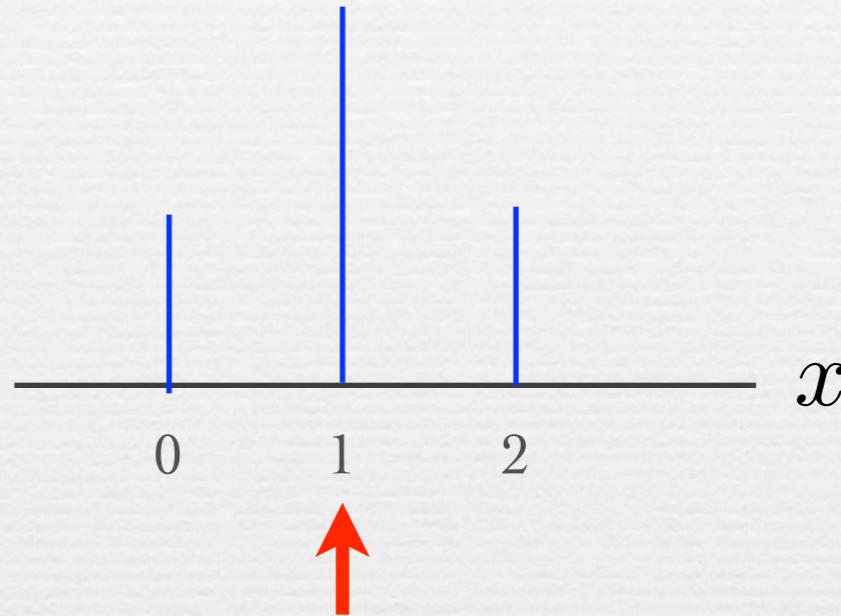
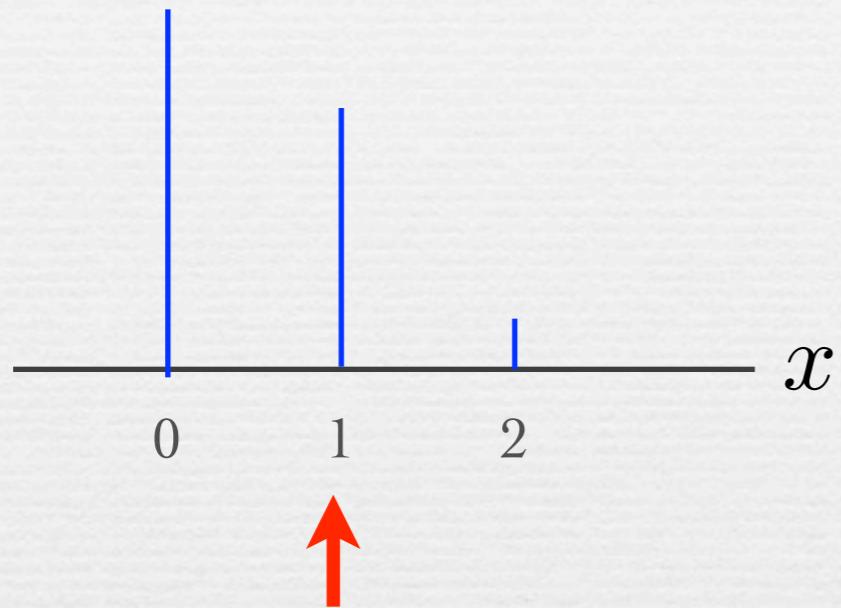
$$P(x|\theta) =$$



But, we don't know θ ...

$$\theta = \frac{1}{4}$$

$$\theta = \frac{1}{2}$$



→ If $x = 1$,

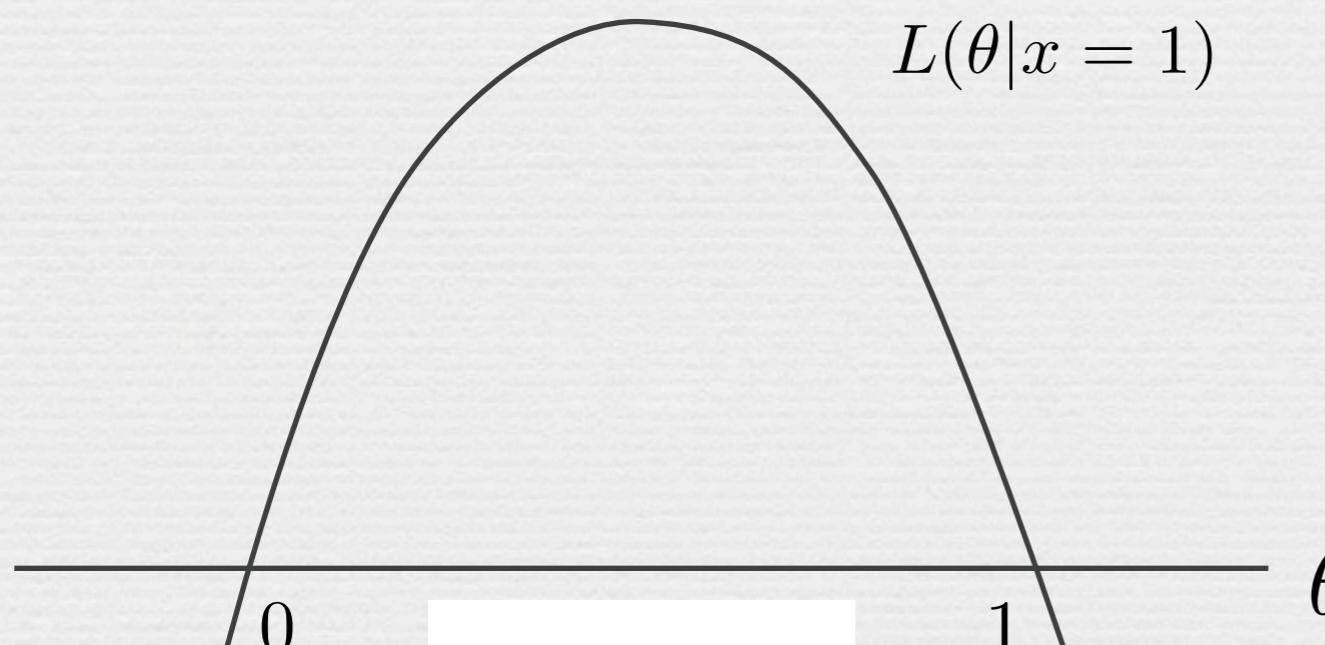
→ Given x ,

→ need to say

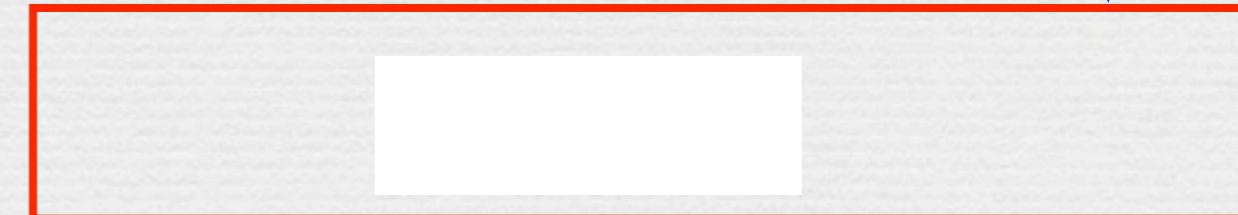
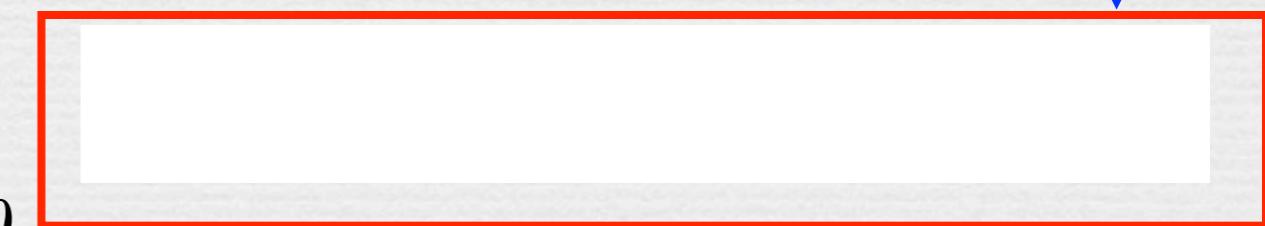
우도함수(likelihood ft.)와 최우추정량 (MLE)

(e.g) 동전 2개를 던져서 앞면이 $x=1$ 개 나왔을 때의 우도 : $L(\theta|x)$

$$L(\theta|x) = \binom{2}{x} \theta^x (1-\theta)^{2-x} \xrightarrow{x=1} L(\theta|x=1) =$$



$L(\theta|x)$ = a function of θ



우도비검정(Likelihood Ratio Test: LRT)

$$H_0 : \theta \in \omega \quad \text{vs.} \quad H_1 : \theta \in \Omega - \omega$$

(e.g.) $H_0 : \mu = 10$ vs. $H_1 : \mu \neq 10$

$\Omega = \text{all real values of } \mu$

$$\sup_{\theta \in \omega} L(\theta) \rightarrow$$

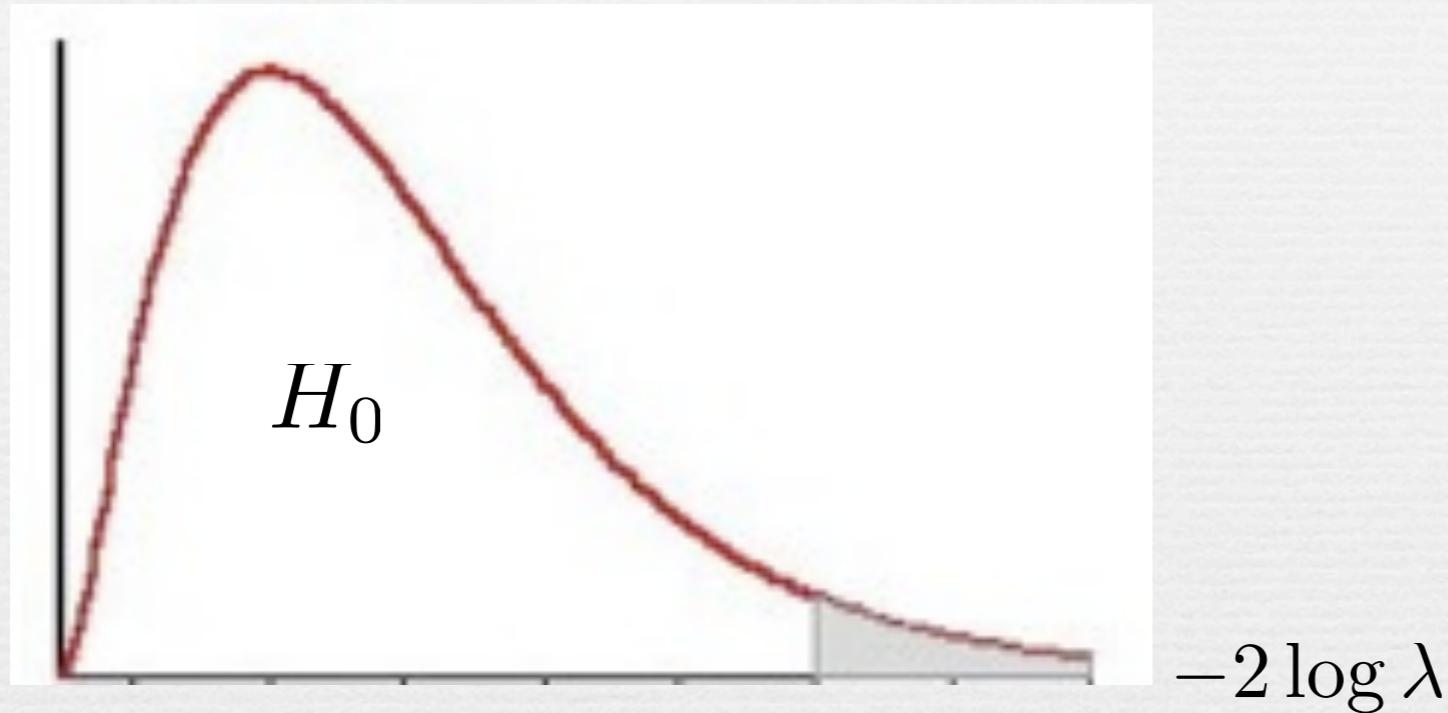
$$\sup_{\theta \in \Omega} L(\theta) \rightarrow$$

LRT 통계량

If H_0 is true \rightarrow

$$\nu = \dim(\Omega) - \dim(\omega)$$

우도비검정(Likelihood Ratio Test: LRT)



If $-2 \log \lambda$

다면량 정규분포의 우도함수 $x_i \sim N_p(\mu, \Sigma)$

우도함수(likelihood function)

$$L(\mu, \Sigma) =$$

로그우도함수(log-likelihood function)

$$l(\mu, \Sigma) = \log L(\mu, \Sigma) = -\frac{n}{2} \log |2\pi\Sigma| - \frac{1}{2} \sum_{i=1}^n (\mathbf{x}_i - \mu)^T \Sigma^{-1} (\mathbf{x}_i - \mu)$$

$$\begin{aligned}\frac{\partial l(\mu, \Sigma)}{\partial \mu} &= 0 \\ \frac{\partial l(\mu, \Sigma)}{\partial \Sigma} &= 0\end{aligned}$$



(S is a sample covariance matrix)

다면량 정규분포와 우도비 검정통계량 LRT

$$H_0 : \boldsymbol{\mu} = \boldsymbol{\mu}_0$$

(e.g.) $H_0 : \boldsymbol{\mu} = \begin{pmatrix} \mu_x \\ \mu_y \end{pmatrix} = \begin{pmatrix} 5 \\ 10 \end{pmatrix} = \boldsymbol{\mu}_0$

$$\sup_{\boldsymbol{\mu} \in \Omega} L(\boldsymbol{\mu}) = L(\boldsymbol{\mu} = \bar{\boldsymbol{x}}, \Sigma = S = \text{[redacted]})$$

$$\sup_{\boldsymbol{\mu} \in \omega} L(\boldsymbol{\mu}) = L(\boldsymbol{\mu} = \boldsymbol{\mu}_0, \Sigma = \text{[redacted]})$$

$$\lambda =$$

$$T_0^2 = n(\bar{\boldsymbol{x}} - \boldsymbol{\mu}_0)^T S^{-1}(\bar{\boldsymbol{x}} - \boldsymbol{\mu}_0)$$

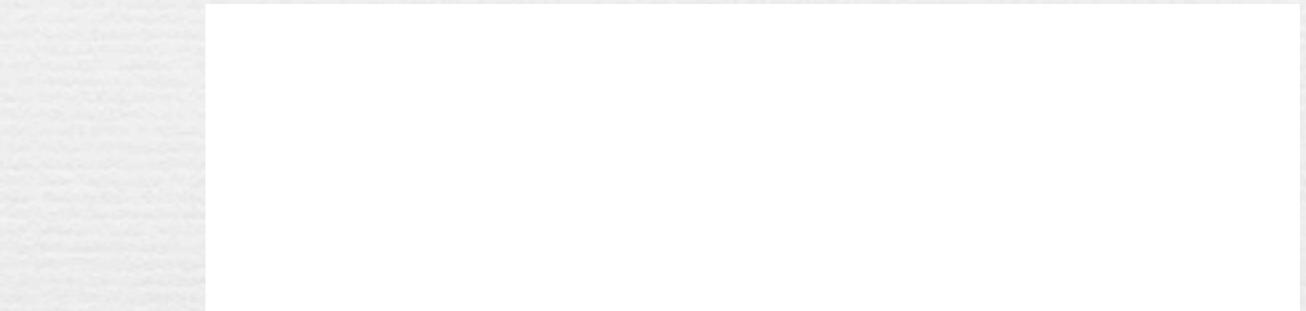
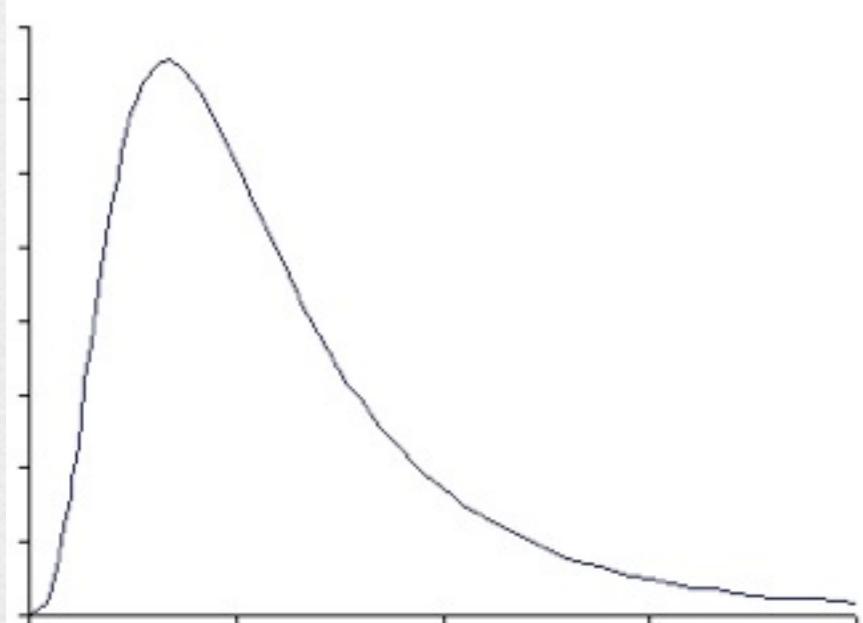


$$\rightarrow \text{If } -2 \log \lambda$$

다변량 정규분포와 Hotelling's T² 통계량

$$H_0 : \mu = \mu_0$$

$$T_0^2 = n(\bar{x} - \mu_0)^T S^{-1}(\bar{x} - \mu_0) \quad \text{호텔링의 } T^2 \text{ 통계량}$$



$$\frac{n-p}{p(n-1)} T_0^2$$

$$\text{If } \frac{n-p}{p(n-1)} T_0^2 >$$

example (sweat data) p.69

$$H_0 : \mu = \begin{pmatrix} \\ \\ \end{pmatrix}$$

- $Y1 = \text{sweat rate} = [(\text{pre-run weight}) - (\text{post-run weight})] * 16 \text{ (oz)}$.. after 1 hr running
- $Y2 = Na$
- $Y3 = Ka$

```

data b; set a;

v1=y1-4; v2=y2-50; v3=y3-10;

proc glm;

model v1 v2 v3 = /nouni;

manova h=INTERCEPT ; run;

```

Manova Test Criteria and Exact F Statistics for
the Hypothesis of no Overall INTERCEPT Effect
H = Type III SS&CP Matrix for INTERCEPT E = Error SS&CP Matrix

S=1 M=0.5 N=7.5

Statistic	Value	F	Num DF	Den DF	Pr > F
Wilks' Lambda	0.66112774	2.9045	3	17	0.0649
Pillai's Trace	0.33887226	2.9045	3	17	0.0649
Hotelling-Lawley Trace	0.51256699	2.9045	3	17	0.0649
Roy's Greatest Root	0.51256699	2.9045	3	17	0.0649

일반선형가정 and Hotelling's T^2

일반선형가정 $H_0 : A\mu = b \longleftrightarrow$

$x_i \sim N_p(\mu, \Sigma) \longrightarrow Ax_i - b \sim$

If H_0 is true,

$$T_0^2 = n(A\bar{x} - b)^T (ASA^T)^{-1}(A\bar{x} - b)$$

(e.g.) $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \sim N_3 \left(\begin{pmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \end{pmatrix}, \begin{pmatrix} \sigma_1^2 & \sigma_{12} & \sigma_{13} \\ \sigma_{21} & \sigma_2^2 & \sigma_{23} \\ \sigma_{31} & \sigma_{32} & \sigma_3^2 \end{pmatrix} \right)$

$$H_0 : \begin{pmatrix} \mu_2 = 12\mu_1 + 2 \\ \mu_2 = 5\mu_3 \end{pmatrix} \longleftrightarrow$$

example (sweat data) p.69

```
data b; set a;  
  
v1=12*y1-y2+2;  
  
v2=y2-5*y3;  
  
proc glm;  
  
model v1 v2 = /nouni;  
  
manova h=INTERCEPT ; run;
```



MANOVA Test Criteria and Exact F Statistics for the Hypothesis of No Overall Intercept Effect					
H = Type III SSCP Matrix for Intercept					
E = Error SSCP Matrix					
	S=1	M=0	N=8		
Statistic		Value	F Value	Num DF	Den DF
Wilks' Lambda		0.67374574	4.36	2	18
Pillai's Trace		0.32625426	4.36	2	18
Hotelling-Lawley Trace		0.48423944	4.36	2	18
Roy's Greatest Root		0.48423944	4.36	2	18