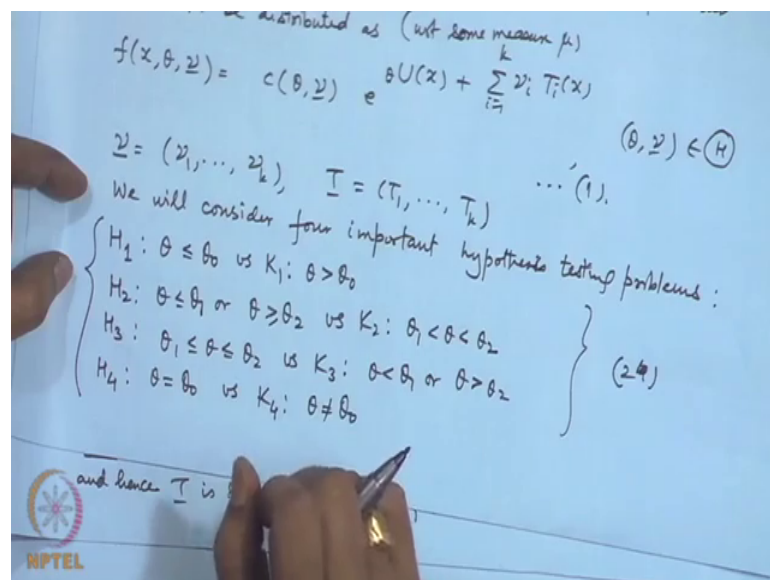


**Statistical Inference**  
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**Lecture – 45**  
**Applications of UMP Unbiased Tests – I**

So, in the previous lecture I have described in detail how for the multiparameter exponential families for the four types of hypothesis testing problems  $H_1$  versus  $K_1$ ,  $H_2$  versus  $K_2$ ,  $H_3$  versus  $K_3$  and  $H_4$  versus  $K_4$  we have UMP unbiased test.

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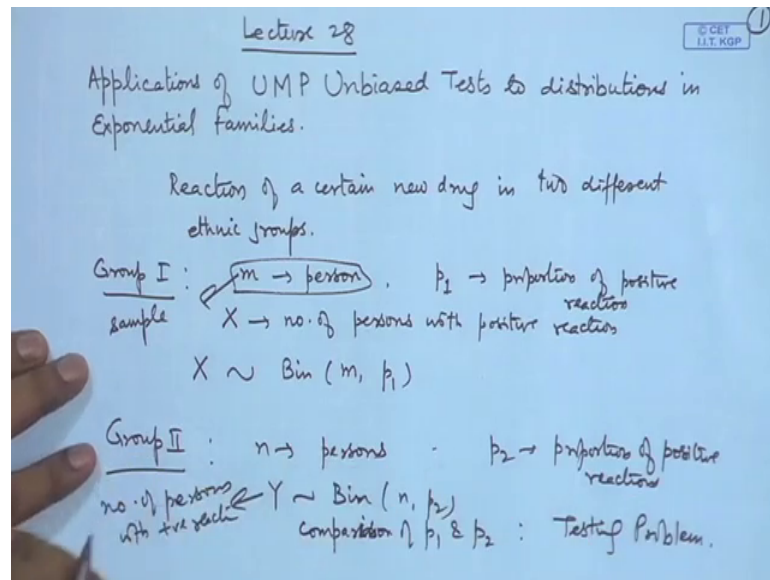


So, the hypothesis testing problems let me show again  $H_1$   $\theta$  less than or equal to  $\theta_0$  against  $\theta > \theta_0$ . So, this is a one sided testing problem if a UMP unbiased test is here for dual problem also it will exist. Now for  $H_2$  we have for  $H_3$  and for  $H_4$   $H_4$  is actually the null hypothesis point hypothesis whereas, the alternative hypothesis is two sided.

For all the cases UMP unbiased test can be found; the method that I described was, firstly, we derived the conditional test and then we derive the unconditional test. The conditional test could have been for  $H_1$  and  $H_2$  UMP test and for  $H_3$  and  $H_4$  there were UMP unbiased, but when we considered unconditional then all the test became UMP unbiased.

So, I will demonstrate certain popular testing problems for example, you are considering parameters of binomial distributions, Poisson's distributions, normal distributions; so these are all you can say popular distributions. Suppose I have two proportions; so, it could be like we are having reaction to certain measure or reaction to certain drug.

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So, let us consider reaction of a certain new drug in two different ethnic groups ok. So, suppose for the first group; group I; we considered that suppose m persons were tested and p 1 is say the proportion of positive reaction say. So, we took the observation and X is the observation number of persons with positive reaction in the sample; if I am considering a sampling of m persons; then here we can write the model as X follows binomial m p 1.

Similarly in the group II, you can consider say n persons are there a sample of n persons is considered and p 2 is the proportion of say positive reactions and then Y follows binomial n p 2; where y is the number of persons out of n with positive reactions. So, the problem could be compare comparison of p 1 and p 2; this could be a testing problem. For example, we may like to check whether in the two populations the people are you can say proportion of the people with positive reactions is the same or not ok. So, we may like to test whether p 1 is equal to p 2 or p 1 is less than p 2 or p 1 is greater than p 2 or simply p 1 is not equal to p 2 etcetera various type of comparison a statements can be given.

Let us see we will demonstrate here that using this procedure that I developed in the last class it is applicable here and we will be able to derive an exact UMP unbiased test for these problems.

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Consider the joint pmf of  $X$  and  $Y$ .

$$f(x, y, p_1, p_2) = \binom{m}{x} p_1^x (1-p_1)^{m-x} \binom{n}{y} p_2^y (1-p_2)^{n-y}$$

$0 < p_1 < 1$   
 $0 < p_2 < 1$

$$= \binom{m}{x} \binom{n}{y} (1-p_1)^m (1-p_2)^n \cdot e^{x \log\left(\frac{p_1}{1-p_1}\right) + y \log\left(\frac{p_2}{1-p_2}\right)}$$

$$= \binom{m}{x} \binom{n}{y} (1-p_1)^m (1-p_2)^n \cdot e^{x \log\left(\frac{p_1}{1-p_1} \cdot \frac{1-p_2}{p_2}\right) + (x+y) \log\left(\frac{p_2}{1-p_2}\right)}$$

$$= h(x, y) c(\theta, \nu) e^{x\theta + (x+y)\nu}$$

where  $\theta = \log\left(\frac{p_1(1-p_2)}{p_2(1-p_1)}\right)$ ,  $\nu = \log\left(\frac{p_2}{1-p_2}\right)$

$\eta = \log p$        $\rho = \frac{p_1(1-p_2)}{p_2(1-p_1)} \rightarrow$  odds ratio.

$U(x, y) = x$   
 $T(x, y) = x + y$

So, consider the joint probability mass function of  $X$  and  $Y$ . So,  $x, y, p_1, p_2$  that is equal to  $m \cdot c \cdot x \cdot p_1$  to the power  $x \cdot 1$  minus  $p_1$  to the power  $m$  minus  $x \cdot n \cdot c \cdot y \cdot p_2$  to the power  $y \cdot 1$  minus  $p_2$  to the power  $n$  minus  $y$ . Here  $p_1, p_2$  both lie between 0 and 1 and this  $x$  and  $y$  values;  $x$  is from 0 to  $m$  and  $y$  is equal to 0 to  $n$ . Now we have written the distribution in this particular fashion it does not look like in the exponential family form. Our aim is to express it in the form of multiparameter exponential family that I described of this nature; we will show it here that it can be done.

So, we write it as say  $m \cdot c \cdot x \cdot n \cdot c \cdot y \cdot 1$  minus  $p_1$  to the power  $m$  minus  $p_2$  to the power  $n$ ;  $e$  to the power  $x \cdot \log\left(\frac{p_1}{1-p_1}\right) + y \cdot \log\left(\frac{p_2}{1-p_2}\right)$ . Now, this you can easily see this is a two parameter exponential family here  $x$  and this is you can say  $\theta$  and this you can say  $\nu$  and it is a that form. However, if I write this particular form I can test either about  $p_1$  or about  $p_2$  whereas, our original  $m$  was to test about a comparison between  $p_1$  and  $p_2$ .

So, here little bit of you can say reparameterization is required what we do we write it in this form;  $m \cdot c \cdot x \cdot n \cdot c \cdot y \cdot 1$  minus  $p_1$  to the power  $m$  minus  $p_2$  to the power  $n$ ;  $e$  to the power  $x \cdot \log\left(\frac{p_1}{1-p_1}\right) + y \cdot \log\left(\frac{p_2}{1-p_2}\right) + x + y$ ;  $\log\left(\frac{p_2}{1-p_2}\right)$

$p_2$ . Let me call this as sum  $h$  of  $x$   $y$ ; this function is  $c$  of our parameters we will rename this thing. So, I am calling it  $\theta$   $\nu$   $e$  to the power  $x$   $\theta$  plus  $x$  plus  $y$   $\nu$  where this  $\theta$  is actually  $\log$  of  $p_1$  into  $1 - p_2$  divided by  $p_2$  into  $1 - p_1$ ;  $\nu$  is equal to  $\log$  of  $p_2$  divided by  $1 - p_2$ . And then this  $U$   $x$  we can take to be  $x$  and  $T$   $x$  we can take to be; so  $U$   $x$   $y$ ;  $T$   $x$   $y$  that is equal to  $x$  plus  $y$ .

We give some names here this quantity; this particular quantity we can call it  $\log$  of sum  $\rho$  for example,  $\rho$  is equal to  $p_1$  into  $1 - p_2$  divided by  $p_2$  into  $1 - p_1$ . So, this  $\rho$  this is also called odds ratio; that means,  $p_1$  by  $1 - p_1$  is odds in favor of events happening with  $x$  and  $p_2$  by  $1 - p_2$  is the odds in favor of events happening with  $y$ .

So, therefore, this ratio is also called odds ratio and we are able to actually write down the parameter  $\theta$  in terms of this. Notice this manipulation is actually helping us to write a function which is involving both  $p_1$  and  $p_2$ . Now let us see whether this is actually helpful in testing a hypothesis regarding comparison between  $p_1$  and  $p_2$ .

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$\theta = 0 \Leftrightarrow p_1 = p_2$        $p = \frac{p_1(1-p_2)}{p_2(1-p_1)} = 1 \Leftrightarrow p_1 = p_2$   
 $\theta \leq 0 \Leftrightarrow p_1 \leq p_2$   
 So we are now able to implement Theorem 2 (given in previous lecture).  
 $U = X, T = X + Y$ .  
 The tests for  $H_1, H_2, H_3, H_4$  will be obtained in terms of conditional dist<sup>n</sup> of  $X$  given  $T = t$ .  

$$P(X = x | T = t) = \frac{P(X = x, X + Y = t)}{P(X + Y = t)}, \quad x = 0, 1, \dots, t$$

$$= \frac{P(X = x, Y = t - x)}{\sum_{x_1=0}^t P(X = x_1, Y = t - x_1)} = \frac{P(X = x) P(Y = t - x)}{\sum_{x_1=0}^t P(X = x_1) P(Y = t - x_1)}$$

Now, let us see if I consider say  $\theta$  is equal to 0  $\theta$  is equal to 0 is equivalent to same  $\rho$  is equal to 1 and  $\rho$  is equal to 1 is equivalent to same  $p_1$  is equal to  $p_2$  because  $\rho$  is this quantity. So, if I say  $\rho$  that is equal to  $p_1$  into  $1 - p_2$  divided by  $p_2$  into  $1 - p_1$  is equal to 1. Then if I write down and expand  $p_1$   $p_2$  will cancel out you will get  $p_1$  is equal to  $p_2$ .

So, if I am considering equality of the two proportions; it is equivalent to hypothesis in terms of my parameter which is appearing here. And if I am looking at the point null hypothesis  $H_4$  that is  $\theta$  is equal to  $\theta_0$ , then  $\theta$  is equal to 0 is the corresponding hypothesis here; let us also see others. If I say  $\theta \leq 0$  then this is equivalent to  $\rho \leq 1$  and this is equivalent to  $p_1 \leq p_2$ . So, the hypothesis of the nature  $H_0, H_1, H_2$  and  $H_3$  can also be written in the form of this.

So, we are now able to implement theorem 2 given in previous lecture. And I have already written that  $U$  is actually equal to  $X$  and  $T$  is equal to  $X + Y$ . The tests for  $H_1, H_2, H_3, H_4$  will be obtained in terms of conditional distribution of  $X$  given  $X + Y = t$ ; so we derive this distribution now. Probability of say  $X = x$  given  $T = t$  that is probability of  $X = x$  given  $X + Y = t$  divided by probability of  $X + Y = t$ ; for  $x = 0, 1, \dots, t$ .

This we can write as probability  $X = x, Y = t - x$  divided by probability  $X = x, Y = t - x$ ; let me write here  $x_1$  and  $x_1 = x$  is equal to  $0$  to  $t$ . Since  $X$  and  $Y$  are independent here this becomes a product here; product of probability  $X = x$  probability  $Y = t - x$  divided by sum of probability  $X = x_1$  probability of  $Y = t - x_1$ ;  $x_1 = 0$  to  $t$ .

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$$\begin{aligned}
 &= \frac{\binom{m}{x} p_1^x (1-p_1)^{m-x} \cdot \binom{n}{t-x} p_2^{t-x} (1-p_2)^{n-t+x}}{\sum_{x_1=0}^t \binom{m}{x_1} p_1^{x_1} (1-p_1)^{m-x_1} \binom{n}{t-x_1} p_2^{t-x_1} (1-p_2)^{n-t+x_1}} \\
 &= \frac{\binom{m}{x} \binom{n}{t-x} p^x}{\sum_{x_1=0}^t \binom{m}{x_1} \binom{n}{t-x_1} p^{x_1}}, \quad x=0, 1, \dots, t \\
 &= \binom{m+n}{t} p^t (1-p)^{m+n-t}
 \end{aligned}$$

$p = \frac{p_1(1-p_2)}{p_2(1-p_1)}$

$\theta \leq 0$  vs  $K_1: \theta > 0$ , use for size condition  $\theta = 0$

then  $P(X=x|T=t) = \frac{\binom{m}{x} \binom{n}{t-x}}{\binom{m+n}{t}}, \quad x=0, 1, \dots, t$   $p=1$

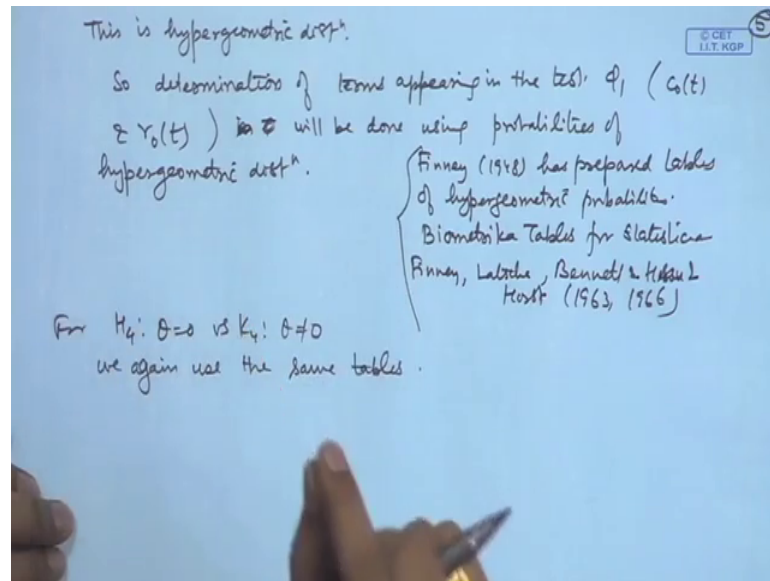
And the distributions of  $X$  and  $Y$  are known; so we can substitute these values and we will get this is equal to  $m \binom{c}{x} p^{x-1}$  to the power  $x-1$  minus  $p^{1-x}$  to the power  $m-x$ ;  $n \binom{c}{t-x} p^{2-x}$  to the power  $p-x$ ,  $1-p^{2-x}$  to the power  $n-t+x$ ;  $p^{x-1}$  to the power  $x-1$  minus  $p^{1-x}$  to the power  $m-x-1$ ,  $n \binom{c}{t-x} p^{x-1}$  to the power  $n-t+x-1$ ;  $x-1$  is equal to  $0$  to  $t$ .

Now, here certain terms will get cancelled out and we can write it in a slightly simplified form. So, we can use the notation that  $\rho$ ;  $\rho$  as  $p^{1-x}$  into  $1-p^{2-x}$  divided by  $p^{2-x}$  into  $1-p^{1-x}$ . So, if we use this notation this can be reduced to  $m \binom{c}{x}$ ;  $n \binom{c}{t-x} \rho^{x-1}$  to the power  $x$  divided by  $\sigma^{x-1}$  is equal to  $0$  to  $t$ ;  $m \binom{c}{x-1}$ ,  $m \binom{c}{t-x-1} \rho^{x-1}$  to the power  $x-1$ ;  $x$  varies from  $0$  to  $t$ .

Now, if you look at this term here  $x-1$  will be is the variable; so, this will be added up actually. So, this becomes finally, a function of  $t$  and  $\rho$ , so this is a function of say  $t$  and  $\rho$  and you are getting a function of  $x$  here; let me call it  $h(x)$  and then  $\rho$  to the power  $x$  here. Suppose we consider say  $H=1$  that is  $\theta$  is equal to  $0$  versus  $\theta$  less than or equal to  $0$  versus  $K=1$   $\theta$  is greater than  $0$ .

So, for size condition  $\theta$  will be equal to  $0$ ; if I take  $\theta$  is equal to  $0$  that is  $\rho$  is equal to  $1$ , then this term will go away. If this term goes away the denominator will be added up as  $m + n \binom{c}{t}$ ; then this probability of  $X$  is equal to  $x$  given  $T$  is equal to  $t$ ; that is becoming simply  $m \binom{c}{x}$ ;  $n \binom{c}{t-x}$  divided by  $m + n \binom{c}{t}$ ; this is for  $x$  equal to  $0$  to  $t$ . And you can notice here that this is nothing, but a hypergeometric distribution; hypergeometric distribution.

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So, determination of terms appearing in the test  $\phi_1$ ; so, let me take up the previous lecture slide we have given the exact form there in the theorem 2. The form of  $\phi_1$  if  $u$  is greater than or equal to  $c$  naught it is  $\gamma$  naught  $u$  is equal to  $c$  naught and 0 if  $u$  is less than  $c$  naught and  $c$  naught and  $\gamma$  naught were determined from this size condition.

Now, here the conditional distribution of  $x$  given  $T$  has been obtained as the hypergeometric distribution; that means,  $c$  naught  $t$  that is  $c$  naught  $t$  and  $\gamma$  naught  $t$  that is in we have given equation number 6, but I do not have to refer it to again. So, this is done using probabilities of hypergeometric distribution; Finney in 1948 has prepared tables of hypergeometric probabilities.

And they have also been published in biometrika tables for statisticians and also by tables of Finney, Latscha, Bennett and Hsu and Horst in 1963, 1966. So, these people have given the tables of the hypergeometric probabilities or you can say the critical points of the hypergeometric distribution which can be used in the determination of this.

For  $H_4$  that is  $\theta$  is equal to 0 versus  $K_4$   $\theta$  naught equal to 0; we again use the same tables that is the distribution that will be used for  $\theta$  is equal to 0; that will again become hypergeometric. So, here we have seen a problem of comparison of proportions in two binomial distributions can be solved using this theory of UMP unbiased tests here.

It is very interesting that we are able to get an exact test; that means, if we are fixing the level of significance then we are having an exact decision making procedure available here whether to accept say  $H_1$  or reject  $H_1$  or accept  $H_4$  or reject  $H_4$  etcetera; that means, whether the proportions are same or one is less than the other etcetera. I will give one more example here.

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2. Suppose we have two <sup>Poisson</sup> arrival processes with arrival rates  $\lambda$  and  $\mu$  respectively. Let  $X$  &  $Y$  be observations from these two processes.  $X \sim P(\lambda)$ ,  $Y \sim P(\mu)$ . We want to compare the arrival rates of the two processes. The joint pmf of  $X$  and  $Y$  is

$$f(x, y, \lambda, \mu) = \frac{e^{-\lambda} \lambda^x}{x!} \cdot \frac{e^{-\mu} \mu^y}{y!}, \quad \begin{matrix} x=0, 1, 2, \dots \\ y=0, 1, 2, \dots \\ \lambda > 0, \mu > 0 \end{matrix}$$

$$= \frac{e^{-(\lambda+\mu)}}{x! y!} e^{x \log \lambda + y \log \mu}$$

$$= \frac{e^{-(\lambda+\mu)}}{x! y!}$$

Suppose we have we have say two arrival processes and they are Poisson arrival prices with arrival rates say  $\lambda$  and  $\mu$  respectively. So, it could be like we are looking at the arrival rates at two service counters; where there are large number of service provider; so we may be looking at those things. We may be looking at the traffic through say a internet service provider we may be looking at the number of packets arriving etcetera there can be various such applications where we may have two arrival processes.

Let  $X$  and  $Y$  be observations from these two processes ok; that means, I am assuming say  $X$  follows Poisson  $\lambda$   $Y$  follows Poisson  $\mu$ ; that means, we have fixed the time or the area etcetera in which this is; that means, they are on the same scales that we are considering. We want to compare the arrival rates of the two processes; so, we will show here again that we can put up the problem in a mutliparameter exponential family and then we can simplify the things.

So, the joint probability mass function of  $X$  and  $Y$  that is equal to  $e$  to the power minus  $\lambda$   $\lambda$  to the power  $x$  by  $x$  factorial;  $e$  to the power minus  $\mu$   $\mu$  to the power  $y$



by  $y$  factorial  $x$  is equal to 0, 1, 2 and so on,  $y$  is equal to 0, 1, 2 and so on;  $\lambda$  and  $\mu$  both are positive. So, this we write as  $e$  to the power minus  $\lambda$  plus  $\mu$  divided by  $x$  factorial  $y$  factorial;  $e$  to the power  $x \log \lambda$  plus  $y \log \mu$ .

Now, this is a two parameter exponential family I can consider this as  $\theta$  and this as  $\mu$  this as  $u$  and this as  $t$ . However, writing down like this does not help me to test about comparison between  $\lambda$  and  $\mu$ ; this will help us to test about  $\lambda$  or testing about  $\mu$ ; however, that is not our problem of interest. So, like in the binomial problem we make some reparameterization or rewriting of this term here.

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Let  $X \sim P(\lambda)$ ,  $Y \sim P(\mu)$   
 We want to compare the arrival rates of the two processes.  
 The joint pmf of  $X$  and  $Y$  is  

$$p(x, y, \lambda, \mu) = \frac{e^{-\lambda} \lambda^x}{x!} \cdot \frac{e^{-\mu} \mu^y}{y!}, \quad x=0, 1, 2, \dots$$

$$y=0, 1, 2, \dots$$

$$\lambda > 0, \mu > 0$$

$$= \frac{e^{-(\lambda+\mu)}}{x! y!} e^{x \log \lambda + y \log \mu}$$

$$= \frac{e^{-(\lambda+\mu)}}{x! y!} e^{x \left[ \log \left( \frac{\lambda}{\mu} \right) + (x+y) \log \mu \right]}$$

So, we write it as  $e$  to the power minus  $\lambda$  plus  $\mu$  divided by  $x$  factorial  $y$  factorial;  $e$  to the power  $x \log \lambda$  plus  $y \log \mu$ ; that means, I have subtracted  $x \log \mu$ ; so I add that thing. So, this becomes  $x \log \lambda$  plus  $y \log \mu$ . So, now you can see this entire thing has become in the same form I can consider this as some  $\theta$ ; this as  $\eta$ , this as  $u$  and this as  $t$ .

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$$= c(\theta, \nu) \lambda^x \mu^y e^{\theta U(x,y) + \nu T(x,y)}$$

$$\theta = \log\left(\frac{\lambda}{\mu}\right), \quad \nu = \log \mu, \quad U = X, \quad T = X + Y.$$

$$\theta = 0 \Leftrightarrow \lambda = \mu \quad \theta < 0 \Leftrightarrow \lambda < \mu$$

$$\theta \leq 0 \Leftrightarrow \lambda \leq \mu \quad \theta > 0 \Leftrightarrow \lambda > \mu.$$

$$H_1: \theta \leq 0 \text{ vs } K_1: \theta > 0 \quad \Leftrightarrow \quad H_1^*: \lambda \leq \mu \quad \text{vs } K_1^*: \lambda > \mu$$

$$H_4: \theta = 0 \text{ vs } K_4: \theta \neq 0 \quad \Leftrightarrow \quad H_4^*: \lambda = \mu \quad \text{vs } K_4^*: \lambda \neq \mu.$$
 UMP unbiased tests for  $H_1$  vs  $K_1$  &  $H_4$  vs  $K_4$  exist based on the conditional dist<sup>n</sup> of  $U$  given  $T=t$ .
 
$$P(X=x | T=t) = \frac{P(X=x, X+Y=t)}{P(X+Y=t)}$$

So, I write the whole thing as  $e$  to the power  $\theta U + \nu T$  into a function of  $\theta$  and  $\nu$  and a function of  $x$  and  $y$ . So, here  $\theta$  is  $\log$  of  $\lambda$  by  $\mu$ ;  $\nu$  is  $\log$  of  $\mu$ .  $U$  is equal to  $X$  and  $T$  is equal to  $X + Y$ . So, if I say  $\theta$  is equal to 0; this is equivalent to saying  $\lambda$  is equal to  $\mu$ . Similarly if I say  $\theta$  is less than or equal to 0; this is equivalent to hypothesis  $\lambda$  is less than or equal to  $\mu$ . If I say  $\theta$  less than 0 that is equivalent to same  $\lambda$  is less than  $\mu$ ; if I say  $\theta$  greater than 0, this is equivalent to saying  $\lambda$  is greater than  $\mu$ .

Therefore we have equivalent of  $H_1, H_2, H_3, H_4$  in terms of  $\theta$  equal to 0,  $\theta$  less than or equal to something,  $\theta$  greater than or equal to something etcetera. In particular if I consider say  $H_1$  that is  $\lambda$  sorry  $\theta$  is less than or equal to 0 versus  $K_1$ ;  $\theta$  is greater than 0 this is equivalent to testing  $\lambda$  is equal to less than or equal to  $\mu$  versus  $\lambda$  is greater than  $\mu$ . And similarly  $H_4$  that is  $\theta$  is equal to 0 versus  $K_4$   $\theta$  is equal not equal to 0; this is equivalent to testing  $H_4^*$ ,  $\lambda$  is equal to  $\mu$  against  $K_4^*$   $\lambda$  is not equal to  $\mu$ .

Now for both of this hypothesis; UMP unbiased tests for  $H_1$  versus  $K_1$  and  $H_4$  versus  $K_4$  exist based on the conditional distribution of  $U$  given  $T$ . So, let us calculate that thing like in the previous problem actually the statements few of the statements will actually be the same  $X$  is equal to  $x$  given  $T$  is equal to  $t$ . So, that is equal to probability  $X$  is equal to  $x$ ;  $X + Y$  is equal to  $t$  divided by probability of  $X + Y$  is equal to  $t$ .

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$$\begin{aligned}
 &= \frac{P(X=x)P(Y=t-x)}{\sum_{x=0}^t P(X=x)P(Y=t-x)}, \quad x=0,1,\dots,t \\
 &= \frac{t! \frac{e^{-\lambda} \lambda^x}{x!} \cdot \frac{e^{-\mu} \mu^{t-x}}{(t-x)! (\lambda+\mu)^t}}{\sum_{x=0}^t \binom{t}{x} \frac{e^{-\lambda} \lambda^x}{x!} \cdot \frac{e^{-\mu} \mu^{t-x}}{(t-x)! (\lambda+\mu)^t}} = \frac{\binom{t}{x} \left(\frac{\lambda}{\lambda+\mu}\right)^x \left(\frac{\mu}{\lambda+\mu}\right)^{t-x}}{\sum_{x=0}^t \binom{t}{x} \left(\frac{\lambda}{\lambda+\mu}\right)^x \left(\frac{\mu}{\lambda+\mu}\right)^{t-x}} \\
 &\quad \downarrow \\
 &\quad 1
 \end{aligned}$$

So  $X|_{X+Y=t} \sim \text{Bin}\left(t, \frac{\lambda}{\lambda+\mu}\right)$ .

So the terms  $g(t), r_0(t), q_1(t), c_2(t), r_1(t), r_2(t)$  in test criteria for  $H_1$  vs  $K_1$  &  $H_2$  vs  $K_2$  can be determined

So, this is equal to as before probability of Y is equal to t minus x divided by probability of; and once again we can write down the full description sigma probability of say X equal to x 1; probability Y is equal to t minus x 1; x 1 is equal to 0 to t. So, here x can go from 0 1 to t only. So, this is e to the power minus lambda lambda to the power x by x factorial; e to the power minus mu, mu to the power t minus x by t minus x factorial; this divided by sigma t c x 1 sorry this is e to the power minus lambda lambda to the power x 1 by x 1 factorial and e to the power minus mu mu to the power t minus x 1 by t minus x 1 factorial x 1 is equal to 0 to t.

Now, here some of the terms will get cancelled out especially e to the power minus lambda plus mu that gets cancelled out in the numerator and the denominator and we can multiply by say t factorial here and t factorial here. So, I get t c x and lambda by lambda plus mu to the power x and mu by lambda plus mu to the power t minus x. What I have done? I have considered division by lambda plus mu to the power t; lambda plus mu to the power t. So, this term can be adjusted here and we get this here and similarly the denominator will give me t c x 1; lambda by lambda plus mu to the power x mu by lambda plus mu to the power t minus x 1 where x 1 is from 0 to t.

Now, if you notice the denominator here this is actually equal to 1 because this is nothing, but a sum of the binomial terms here you are getting something like p to the power x and q to the power n minus x here and sigma m c x from 0 to a. So, this term is

one; so this is actually reducing to a binomial distribution. So, the conditional distribution of  $X$  given  $X$  plus  $Y$ ; this is binomial with parameter  $t$  and  $\lambda$  by  $\lambda$  plus  $\mu$ .

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In this conditional situation there exists a UMP test for testing  $H_1$  vs  $K_1$  with test fn  $\phi_1$  given by

$$\phi_1(u, \underline{t}) = \begin{cases} 1 & u > c_0(\underline{t}) \\ \gamma_0(\underline{t}) & u = c_0(\underline{t}) \\ 0 & u < c_0(\underline{t}) \end{cases} \dots (5)$$

where  $c_0(\underline{t})$  &  $\gamma_0(\underline{t})$  are determined by the size condition

$$E_{\theta_0}(\phi_1(U, \underline{I}) | \underline{I} = \underline{t}) = \alpha \quad \forall \underline{t} \dots (6)$$

Similarly  $\exists$  UMP test  $\phi_2$  for testing  $H_2$  vs  $K_2$  given by

$$\phi_2(u, \underline{t}) = \begin{cases} 1, & c_1(\underline{t}) < u < c_2(\underline{t}) \\ \gamma_i(\underline{t}), & u = c_i(\underline{t}), i=1,2 \\ 0, & u < c_1(\underline{t}), \text{ or } u > c_2(\underline{t}) \end{cases} \dots (7)$$

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there exists a UMP test for testing

and  $c_i(\underline{t})$  &  $\gamma_i(\underline{t})$  are determined by

$$E_{\theta_0}(\phi_4(U, \underline{I}) | \underline{I} = \underline{t}) = \alpha \dots (11)$$

$$E_{\theta_0}(\bigcup \phi_4(U, \underline{I}) | \underline{I} = \underline{t}) = \alpha \quad E_{\theta_0}(U | \underline{I} = \underline{t}) \dots (12)$$

We have interpreted the test fns.  $\phi_1, \phi_2, \phi_3, \phi_4$  as conditional tests given  $\underline{I} = \underline{t}$ . Reinterpret them as dependent on  $(U, \underline{I})$ , we have the following theorem.

**Theorem 2:** The test functions  $\phi_1, \phi_2, \phi_3, \phi_4$  are UMP unbiased for test  $H_1$  vs  $K_1, H_2$  vs  $K_2, H_3$  vs  $K_3$  and  $H_4$  vs  $K_4$  respectively.

Therefore, in the test function if I am considering  $c$  naught and  $\gamma$  naught here from this condition or in  $\phi_4$  function if we are considering these coefficients that is  $c_1, c_2$ ;  $\gamma_1, \gamma_2$  etcetera through these conditions they all can be determined from the tables of the binomial distribution.

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$E_{\theta} \{ \phi_2(U, T) | T = t \} = \alpha, \quad i=1, 2. \quad \dots (8)$   
 For  $H_3$  vs  $K_3$ , UMP unbiased test  $\phi_3$  is given by  

$$\phi_3(u, t) = \begin{cases} 1, & u < c_1(t) \text{ or } u > c_2(t) \\ r_i(t), & u = c_i(t), \quad i=1, 2 \\ 0, & c_1(t) < u < c_2(t) \end{cases} \quad \dots (9)$$
 where  $c_1(t), r_i(t)$  are determined by  
 $E_{\theta} \{ \phi_3(U, T) | T = t \} = \alpha, \quad i=1, 2. \quad \dots (10)$   
 UMP unbiased test  $\phi_4$  is given by  
 $\phi_4(u, t)$

So, the terms  $c_1, c_2, \gamma_1, \gamma_2$  in test criteria for  $H_1$  versus  $K_1$  and  $H_4$  versus  $K_4$  can be determined from the tables of binomial probabilities.

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from the tables of binomial probabilities  
 In fact for  $\lambda = \mu$ .  $\Leftrightarrow \theta = 0$   
 $X | X+Y=t \sim \text{Bin}(t, \frac{1}{2})$  (So this is quite simple)

In fact, when we are considering the point  $\lambda$  is equal to  $\mu$ ; then that case that is  $\theta$  is equal to 0 then  $X$  given  $X + Y$  is equal to  $t$  this is simply following binomial  $t$  half. So, this is simply it can be calculated directly without even seeing the tables here; so this is quite simple.

I have shown you two applications of the UMP unbiased test, where we are considering two different populations. Like in the case of binomial we had two proportions  $p_1$  and  $p_2$ ; in Poisson arrival we have two rates  $\lambda$  and  $\mu$  and we were able to compare. We will further show the applications to comparing the means etcetera for normal populations and variances; also testing for the means and variances in normal population.

So, these are applications I will be; so I will be covering it in the next lecture.