

Engineering Statistics
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Lecture 45
Calculations of Power Functions

(Refer Slide Time: 0:27)

Error in Hypothesis Testing

$H_0 : \theta \in \Theta_0$ null hypothesis
 $H_1 : \theta \in \Theta_0^c$ alternate hypothesis

- ▶ Rejection set $\mathcal{R} = \{x : \lambda(x) \leq c\}$. If $x \in \mathcal{R}$ hypothesis H_0 is rejected, otherwise H_1 is accepted
- ▶ In accepting or rejecting hypothesis the experimenter could be making mistake
- ▶ How to control the error?

Types of Error $\mathcal{R}_c = \{x : \lambda(x) \leq c\}$

There could be two types of error

- ▶ Type I error: If $\theta \in \Theta_0$ (H_0 is true), but the hypothesis test incorrectly **rejects** the null hypothesis H_0
- ▶ Type II error: If $\theta \in \Theta_0^c$ (H_1 is true), but the hypothesis test incorrectly **accepts** the null hypothesis H_0

Truth/Decision	Accept	Reject
$\leftarrow H_0$	Correct ✓	Type I error
$\rightarrow H_1$	Type II error	Correct ✓

$H_0 :$
 $H_1 :$

Power Function

Given \mathcal{R}

- If $\theta \in \Theta_0$, $P_\theta(\mathbf{X} \in \mathcal{R})$ gives probability of Type I error
- If $\theta \in \Theta_0^c$, $1 - P_\theta(\mathbf{X} \in \mathcal{R})$ gives probability of Type II error.

Power Function of a hypothesis test with rejection region \mathcal{R} is a function of θ given by $\beta(\theta) = P_\theta(\mathbf{X} \in \mathcal{R})$

For an 'ideal' hypothesis test

- What should be the value of $P_\theta(\mathbf{X} \in \mathcal{R})$ if $\theta \in \Theta_0$?
- What should be the value of $P_\theta(\mathbf{X} \in \mathcal{R})$ if $\theta \in \Theta_0^c$?
- The ideal case cannot be attained in practical problems
- A good hypothesis test should be such that $P_\theta(\mathbf{X} \in \mathcal{R}) \rightarrow 0$ when $\theta \in \Theta_0$ and $P_\theta(\mathbf{X} \in \mathcal{R}) \rightarrow 1$ when $\theta \in \Theta_0^c$

$\beta(\mathbf{X} \in \mathcal{R})$ is Type-I $\theta \in \Theta_0$
 $\beta(\mathbf{X} \in \mathcal{R}^c)$ is Type-II $\theta \in \Theta_0^c$
 $\mathbf{X} \sim f(\cdot|\theta)$

Type I vs Type II error

Example: $X \in \text{Bin}(5, \theta)$

$H_0: \theta \leq 1/2$ null hypothesis
 $H_1: \theta > 1/2$ alternate hypothesis

$\mathcal{R} = \{(1, 1, 1, 1, 1)\}$

$\beta(\theta) = P_\theta(\mathbf{X} \in \mathcal{R}) = \theta^5$

Test 1: Reject H_0 iff all success are observed

Type I error: $\beta_1(\theta) \leq (1/2)^5 = 0.0312$ for all $\theta \leq 1/2$
 Type II error: $1 - \beta_1(\theta) > 0.97$

Test 2: Reject H_0 if 3, 4, 5 success are observed

$\beta_2(\theta) = \binom{5}{3} \theta^3 (1-\theta)^2 + \binom{5}{4} \theta^4 (1-\theta) + \theta^5$

$\beta(\theta) \leq (1/2)^5$ $\theta \leq 1/2$
 upper bound on Type-I error.

Example Contd.

Suppose we want Type I error to be at most 0.1 and Type II error to be at most 0.2 when $\theta > \theta_0 + \sigma$. What should be the value of c and n ?

$$\beta(\theta) = P\left(Z \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}}\right)$$

- Increasing in θ
- set $\beta(\theta_0) = 0.1$ and $\beta(\theta_0 + \sigma) = 0.2$
- For a fixed sample, it is impossible to make both types of errors arbitrarily small.
- To find a good test, it is common to restrict Type I at a fixed level.

In the last class we started talking about this hypothesis testing. We introduced what is hypothesis testing, and we defined what is a rejection set, we talked about various types of error particularly we talked about type 1 error type 2 error, and we defined type 1 and type 2 error using your Power Functions, and then discussed what should be the ideal hypothesis should look like in terms of its properties for power function, and then we discussed some examples especially about this binomial case and Gaussian case.

(Refer Slide Time: 1:28)

Type I vs Type II error LRT : $R = \{x: X_1 \leq c\}$

Example: $X \in \text{Bin}(5, \theta)$

$H_0: \theta \leq 1/2$ null hypothesis
 $H_1: \theta > 1/2$ alternate hypothesis

$R = \{(1, 1, 1, 1, 1)\}$
 $P_\theta(X \in R) = \theta^5$
 $\beta(\theta) = P(X_1=1, X_2=1, \dots, X_5=1) = \theta^5$

▶ Test 1: Reject H_0 iff all success are observed
 $\beta_1(\theta) = \Pr(\{1, 1, 1, 1, 1\}) = \theta^5$

Type I error: $\beta_1(\theta) < (1/2)^5 = 0.0312$ for all $\theta \leq 1/2$
 Type II error: $1 - \beta_1(\theta) > 0.97$

▶ Test 2: Reject H_0 if 3, 4, 5 success are observed
 $\beta_2(\theta) = \binom{5}{3} \theta^3(1-\theta)^2 + \binom{5}{4} \theta^4(1-\theta) + \theta^5$

Power Function Given R

▶ If $\theta \in \Theta_0$, $P_\theta(X \in R)$ gives probability of Type I error
 ▶ If $\theta \in \Theta_0^c$, $1 - P_\theta(X \in R)$ gives probability of Type II error.

Power Function of a hypothesis test with rejection region R is a function of θ given by $\beta(\theta) = P_\theta(X \in R)$

$P_\theta(X \in R)$ is Type-I
 $\theta \in \Theta_0$
 $P_\theta(X \in R^c)$ is Type-II
 $\theta \in \Theta_0^c$

$X \sim f(\theta)$

For an 'ideal' hypothesis test

- ▶ What should be the value of $P_\theta(X \in R)$ if $\theta \in \Theta_0$?
- ▶ What should be the value of $P_\theta(X \in R)$ if $\theta \in \Theta_0^c$?
- ▶ The ideal case cannot be attained in practical problems
- ▶ A good hypothesis test should be such that $P_\theta(X \in R) \rightarrow 0$ when $\theta \in \Theta_0$ and $P_\theta(X \in R) \rightarrow 1$ when $\theta \in \Theta_0^c$

So, what we will do today is little bit discuss this example more. So, we said let us say we have a sample coming from a binomial distribution which has parameter 5 and theta. We could equivalently think of these are like a 5 samples coming from a Bernoulli random variable with parameter theta. But here I am just taking one sample coming from a binomial

which I could... Now we want to check whether the θ is taking value less than half or it is taking value more than half, we had to decide a test here. One test like you can always go with the LRT test. Let us discuss two other possible test.

Now, since we are discussing whether θ is less than half and more than half, one natural test we can think of is when more number of 1s are in fact, all are 1s maybe that is θ greater than half otherwise, it is less than half that is one possible test. And that is what let us define my region are that is the region rejection region as something simply 1, all these 1s. So, when all 1s I see I will reject it to be a null hypothesis. Otherwise, I accept it to be a null hypothesis, so this is my rejection region had just one vector in that.

Now if I had to calculate the probability that under my θ , what is that x belongs to R ? Now X is I mean binomial but I could think of like it is a Bernoulli with 5 samples. So, this is nothing but we discussed last time, this is nothing but simply probability that X_1 is 1, X_2 is 2, all sorry, X_2 is 2 all the way up to X_5 is 1 all of them has to be 1 which is because of the independent nature, all of them it happened probability θ , so they select that θ to the power 5.

Now let us compute type 1 error for this. What is the definition of type 1 error? Definition of, so what we basically got, is this is the definition of our power function. What how did we denote the power function? We denote it as β of θ . So, this is like a β of θ . So, let us plot it my θ is going to take value obviously between 0 to 1 and this is my β of θ . Now, how does θ to the power 5 look like?

That is some polynomial curve let us say it starts from 0 and goes till 1. Let us say, this is like something looks like this and let us say this is the half value and this is what my θ not is this θ not is actually half for me. So, when θ is less than half and I compute this probability that β of θ what this will give me this will actually give me a type 1 error. In this region because θ is less than half that is corresponding to the null hypothesis and I am rejecting the samples that is my type 1 error.

So, in this region it is giving me type 1 error. There are what is this is giving in this region. Now, θ here is greater than half that means it is coming from alternative hypothesis, but this probability is still that of rejection. So, what I am basically doing is I am θ is now in the null hypothesis part, sorry alternate hypothesis, but I am actually still plotting this only not 1 minus. So, this is actually given me the compliment of the error probability. So, this is

like so, in this portion this is type 1 error, sorry in this portion it is type 1 error and in this portion it is 1 minus type 2 error.

Now, we know that this function is increasing in theta, what is the maximum value of type 1 error here is. So, the type 1 error is happening here the maximum value of type 1 error happens when theta is equals to half and that is why we are written maximum type 1 error is half to the power 5, that is 0.0312. Now, let us look into the type 2 error. So, what is the type 2 error now? What will be its... So, let us try to actually compute its smallest value.

Now, 1 minus this is going to give you 1 minus type 2 error is going to give you or that will be represented by this region. Now, maybe just if I look into the complement of this maybe it will go something like this I am just like trying to do the complement of this. This is like this is basically 1 minus theta to the power 5. Beta 1 theta is anyway this is defined for every theta and that is theta to the power 5. And since beta 1 of theta is increasing in theta if, when theta becomes largest value 1, then it will take the smallest value that is going to be 0, that means there will not be any type 2 error if theta is already 1.

But now, the smallest value is going to happen for this when theta is going to be half and when I plug half here, that is basically this value you will see that you are going to get this is actually we saw that last time this is kind of approximately 0.97. So, you see that type 1 error is very small that you rejecting a sample which is coming from a null hypothesis is very small. On the other hand, type 2 error also very high. It may happen that you were sample is coming from alternate hypothesis, but you may end up accepting it more very good amount of time. But it is a bad case for us.

So, one this is happening because we are looking at a very bad test, we are asking that when everybody is one, that time only I am going to reject otherwise, I am going to accept it to be null hypothesis. So, you can relax this test and say that instead of all to be 1 when the majority are 1, maybe I will reject otherwise I will still accept. So, in that case, we are going to reject only when there are 3 4 or 5 successes and you can compute the corresponding power function for that to be like this. And it so happens that you can also plot this function and this will something turn out to be like this.

This is for a beta 1 and this is for beta 2. So, notice that in this region, which is giving me that type 1 error beta 1 the test one is good, because it has a low 1 here, but it was taking hit in the type 2 error. On the other hand, if you now look into the beta 2, beta 2 has the higher type 1

error compared to, sorry, this type test 2 has a higher type 1 error compared to test one, but whatever it is type 2 error. So, type 2 error is the complement of this. So, since this large compared to this that type error is going to be lower.

So, there is always a tradeoff. It is not that one test is going to give you a small type 1 error, it is not necessarily that it also gives you small type 2 error. There may be other test, which may give you smaller type 2 error, but it may end up having a type 1 error. So, we would always... but what you like ideally, you want both type 1 and type 2 error to be small that would be the best case. So, but that may not always happen simultaneously reducing them may not be possible. So, we have to keep that in mind whenever we are designing a test.

(Refer Slide Time: 13:05)

Example:

Let $\mathbf{X} = \{X_1, X_2, \dots, X_n\}$ is iid drawn from population $\mathcal{N}(\theta, \sigma^2)$ with known σ^2 . We want to test the hypothesis

$H_0: \theta \leq \theta_0$ null hypothesis
 $H_1: \theta > \theta_0$ alternate hypothesis

$\beta(\theta) = P(\mathbf{x} \in \mathcal{R})$

$$\mathcal{R} = \{\mathbf{x} : \lambda(\mathbf{x}) \leq c\} = \left\{ \mathbf{x} : \frac{\bar{x} - \theta_0}{\sigma/\sqrt{n}} \geq c \right\}$$

$$\beta(\theta) = P\left(\frac{\bar{x} - \theta_0}{\sigma/\sqrt{n}} \geq c \right)$$

$$= P\left(\frac{\bar{x} - \theta}{\sigma/\sqrt{n}} \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}} \right)$$

$$= P\left(Z \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}} \right)$$

where $Z \sim \mathcal{N}(0, 1)$.

So, we discussed this again, I think last time we did the computation of this, I am not going to discuss this again, when we have the samples coming from a Gaussian distribution with parameter theta and sigma square and where theta is unknown, but sigma square is known, we can come up with our rejection region through this condition. I hope all of you followed the derivation of the last time, this is for a given c, c something parameter that has been given to you. Now, I can define. Now I know my rejection region, I can define my power function for a given theta.

So, the power function at point theta simply probability that this ratio is going to be larger than c that is coming from. Everybody agree till this point, basically, we know that probability of theta is probability that my x belongs to the rejection region and this rejection region is captured by this that is why we got. Now I can do a simple manipulation here. What

I can do is on this left side, I simply add theta and minus theta and I returned one theta here and the remaining theta minus theta naught minus theta I moved to the other side. Why I had to do this?

Now, if you notice that so by the way, this probability under the parameter theta that is the definition said us. Now what is this quantity now? x bar, now all the samples here are coming from PDF with parameter what? So the samples that I am going to deal with here, when I am computing this probability, they are coming with a PDF parameter theta that is the meaning of this notation.

Now so, what will be the expectation of x bar? It is going to be theta what is x bar? Sample mean and so that basically have centralized and sigma by square root of n. I have normalized it. So, what this random variable is going to be, so, we have seen this much before also this is going to be a normal distribution and that is why I represented it by Z where that is a normal distribution or Gaussian with parameters 01 and that being larger than this. And I know how to compute this probability for a given c theta naught theta sigma and the square root n I know how to compute this probability. Any questions so far on this computations?

Now, I said in the last example both type 1 error type 2 error simultaneously you cannot play because we should try to reduce one other may increase. So, often we may be specified some value I want this much of type 1 error and this match of type 2 error. How to guarantee that then?

(Refer Slide Time: 17:18)

Unbiased Test

A test with power function $\beta(\theta)$ is unbiased if we have $\beta(\theta') \geq \beta(\theta'')$ for any $\theta' \in \Theta_0^c$ and $\theta'' \in \Theta_0$

Example: Let $\mathbf{X} = \{X_1, X_2, \dots, X_n\}$ is iid drawn from population $\mathcal{N}(\theta, \sigma^2)$ with known σ^2 . We want to test the hypothesis

$H_0 : \theta \leq \theta_0$ null hypothesis
 $H_1 : \theta > \theta_0$ alternate hypothesis

$\beta(\theta) = P\left(Z \geq \frac{\theta_0 - \theta}{\sigma/\sqrt{n}}\right)$

Since $\beta(\theta)$ is increasing. It is clear that $\beta(\theta)$ is unbiased.

Now, recall that your power function depends on multiple factors one is the c that you are going to use in your test right, what is the threshold we are going to use in your likelihood ratio test, and also how many samples that you are going to use in performing this test, and of course, what is the threshold you are using to define your null hypothesis and the alternate hypothesis and θ is of course your parameter in this function.

So, if I am going to change the c and n obviously, my power function value is going to change which is in turn is going to change your type 1 and type 2 error. So, somehow if somebody asked you I want this much of type 1 error and type 2 error then how is it possible for you to give him that kind of type 1 and type 2 error? So, what is in control what is in your control here through what you can control our change you were type 1 and type 2 error here. See, if you change c your type 1 and type 2 error will change maybe that is one thing you have in your control and n number of samples that is another thing which in your control by playing with it you can again control type 1 error.

(Refer Slide Time: 18:59)

Example Contd.

Suppose we want Type I error to be at most 0.1 and Type II error to be at most 0.2 when $\theta > \theta_0 + \sigma$. What should be the value of c and n ?

Handwritten notes:
 max value? Type-I error
 $\beta(\theta_0) = P(Z \geq c) = 0.1$
 $1 - \beta(\theta) = 0.2$
 $1 - P(Z > c - \sqrt{n}) = 0.2$
 $1 - P(Z > \infty) = P(Z > \infty) = 0$
 $1 - P(Z > -\infty) = P(Z > -\infty) = 1$
 $1 - P(Z > c - \sqrt{n}) = P(Z > \infty) = 0$
 $1 - P(Z > -\infty) = P(Z > -\infty) = 1$

$$\beta(\theta) = P\left(Z \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}}\right)$$

Handwritten notes:
 $\theta \rightarrow \infty$
 $\theta \rightarrow -\infty$
 $1 - P(Z > c - \sqrt{n}) = P(Z > \infty) = 0$
 $1 - P(Z > -\infty) = P(Z > -\infty) = 1$

- Increasing in θ
- set $\beta(\theta_0) = 0.1$ and $\beta(\theta_0 + \sigma) = 0.2$
- For a fixed sample, it is impossible to make both types of errors arbitrarily small.
- To find a good test, it is common to restrict Type I at a fixed level.

Handwritten notes:
 give value? c .
 Type-I
 1 - Type II

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Example Contd.

Suppose we want Type I error to be at most 0.1 and Type II error to be at most 0.2 when $\theta > \theta_0 + \sigma$. What should be the value of c and n ?

max value? Type-I error
 $\beta(\theta_0) = P(Z \geq c) = 0.1$

min test Type-II
 $\beta(\theta) = P(Z \geq c + \frac{\theta_0 - \theta}{\sigma/\sqrt{n}})$
 $\text{Type-I} \leq \alpha$

1 - P(Z) = 0.2
 $1 - P(Z > c - 1\sigma) = 0.2$

- Increasing in θ
- set $\beta(\theta_0) = 0.1$ and $\beta(\theta_0 + \sigma) = 0.2$
give value? c.
- For a fixed sample, it is impossible to make both types of errors arbitrarily small.
- To find a good test, it is common to restrict Type I at a fixed level.

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Now, if somebody is asking you I want this much of type 1 and type 2 error maybe then you can see what is that c and n that can give you that much that kind of a type 1 and type 2 error let us look into that. Suppose somebody asked you see I want type 1 error to be at most 0.1. And also a type 2 error to be at most 0.2 then my θ is going to be greater than $\theta_0 + \sigma$. When this sum condition you will see why this is coming.

Then somebody is asking you already I want this much of type 1 error and this much of type 2 error. So, the only thing that you have in your control is c and n . Now, you need to decide what c I have to use, what n I have to use, so that I can guarantee this. Let us try to see how can I choose my c and n . So, let us get started with this power function we have, this I hope from the previous slide, it is all clear to you.

Now, let us see this is a let us try to look into this, a little more carefully as a function of θ . So, recall Z is a standard normal distribution, if I let θ here, this θ go to plus infinity, what will be this probability value everybody see agree? 1, are you sure? So, as θ goes to minus infinity, this entire quantity becomes infinity, minus infinity basically. And it does not matter what is the value of c , because this guy dominates the c irrespective of what is this guy is going to be minus infinity and we know that probability that is greater than minus infinity is 1. On the other hand, what is the value if I let θ go to minus infinity, it is going to be 0, everybody agree?

So, as I go from minus infinity to plus infinity, am I going from 0 to 1 in a linear fashion or like in an increasing fashion? Yes, so, if I put this θ , so, as θ goes to infinity, this is going to be 1 and as θ is minus infinity is this is like something like this as a function of

theta, notice that this is not a CDF function. I am plotting it as a function of theta now. This is like 1. Now based on this understanding, can we now decide what should be the value of c and n , see first it is saying first one is type 1 error to be at most 0.1.

So, the maximum value of the type 1 error should be 0.1. Now, can you tell me what is the maximum value of type 1 error here and notice that theta, this is my theta and here it is below this, it is type 1 and above this, this is 1 minus type 2 error. Now, what will be the maximum value of type 1? So, this is where my type 1 error is. And this type 1 error is happening at because to this. So, my maximum value of type 1 error is actually at beta equals to theta what is this value probability that, z is greater or equal to c and which you have been told you to be set to what value 0.1.

Now, from this can you find out what is the value of c , how is that. So, you know that z has to be greater than or equals to c is 0.1 you know exactly at what point your standard normal is going to take value 0.1 like at what point if it exceeds it will take value of 0.1 you can find that. So, you find out now, value of c so, this will give you have so far figured out c , is there something else you need to figure out, n also you need to figure out how to do that n now.

So, for that, let us do the second condition that you are type 2 error has to be at most 0.2. So, I said type 2 error is like a complement of this. It can be like something take me just complement of this when theta is going to be greater than equal to theta. Now, this quantity has to be at most 0.2 when theta is going to be larger than theta plus now, if I look into a type 2 error, is type 2 error is going to be increasing or decreasing function in theta.

So, type 2 error is now going to fall from 1 minus beta theta and we know that if beta of theta is increasing 1 minus beta of theta has to decrease and its maximum value happens when theta equals to theta naught. But, we have been told that when theta is greater than theta not plus sigma that time this value should be at most 0.2. So, you just plug in theta equals 2 theta not plus sigma you will get. So, what you want is I want this quantity at theta to be at theta not plus sigma to be 0.2. And now, let us see what is beta of theta plus sigma if you plug in?

If you plug in here this is like a theta right this will knock off this theta I will get minus sigma that will cancel with this. So, I will get probability that z is greater than or equals to c minus square root n to be... Do I get this condition or maybe let me write this like properly. Now, p minus z is greater than or equal to c minus square root n equals to 0.2 and see you have already decided from this so, plugging that c and now, what is remaining is only n is the

quantity that is remaining and again by using your tail properties of your CDF of your standard normal distribution, you should be able to find out what is n . So, if somebody has to ask you to control your type 1 and type 2 error, the thing that is in your control is what is the c value you can choose and how many samples that you want to guarantee that type 1 and type 2 errors.

So, if you fix a number of sample, maybe it is almost not possible or almost impossible to make both type 1 and type 2 errors to be small. That is why often in practice, what we do is we say that, let us my primary criteria is going to be on type 1 error, let us guarantee that type 1 error is always going to be less than this much. And after that, see that what is the smallest type 2 error you can get?

So, in a way, if you want to look this into the from the optimization point of view what you want to do is you have been asked to do type 1 error to be let us say some quantity, let us call that to be α . And now what you want to do is minimize your type 2 error. So, you want to find a test so this is across all tests. So, for every test is going to be associated with type 1 and type 2 error. And you have been already told type 2 error cannot be that is my hard constraint. Type 1 error cannot be more than α . If you satisfy that is good, but I will be also more happy if you make my type 2 error also small.