

Errors in Hypothesis Testing

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Example

- We are interested in burning rate of a solid propellant used to power aircrew escape systems
- Burning rate is a random variable that can be described by a probability distribution
- Suppose our interest focus on mean burning rate
- $H_0: \mu = 50$ centimeters per second
- $H_1: \mu \neq 50$ centimeters per second

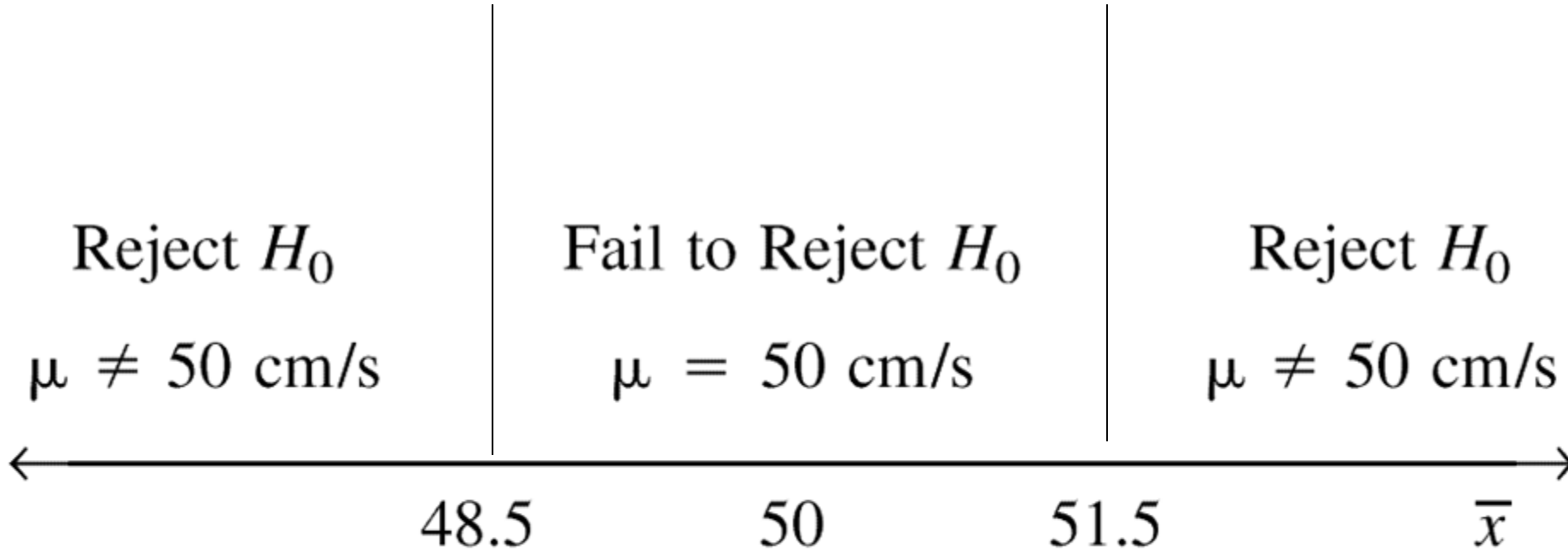


Reference: Applied statistics and probability for engineers, Douglas C. Montgomery, George C. Runger, John Wiley & Sons, 2007

Value of the null hypothesis

- The value of the null hypothesis can be obtained by
 - Past experience or knowledge of the process, or even from the previous tests or experiments
 - From some theory or model regarding the process under study
 - From external consideration, such as design or engineering specifications, or from contractual obligations

Note: for this example $n=10$



Decision criteria for testing $H_0: \mu = 50 \text{ cm/s}$ versus $H_1: \mu \neq 50 \text{ cm/s}$.

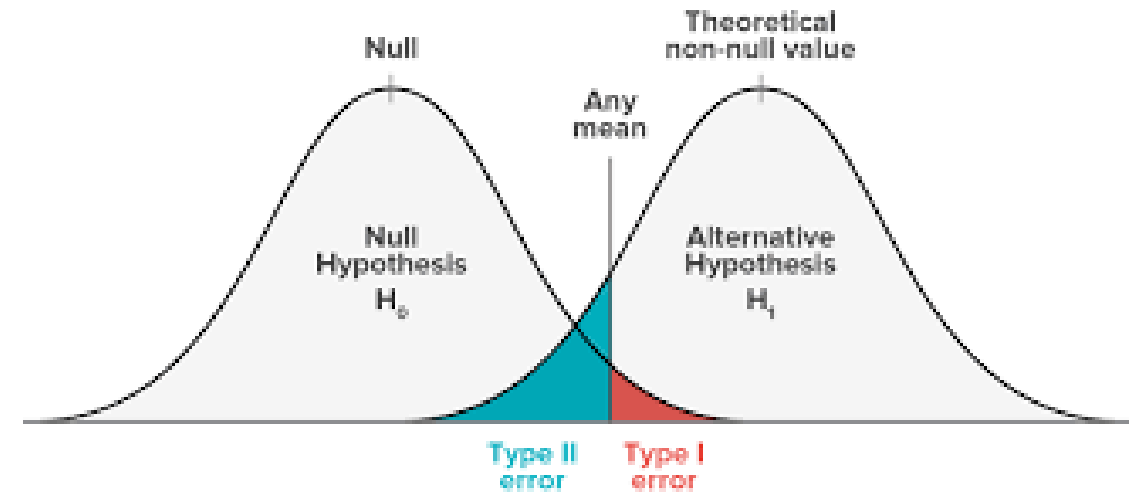
Note: for this example we will
assume $\sigma = 2.5$

Type I Error

- The true mean burning rate of the propellant could be equal to 50 centimeters per second
- However randomly selected propellant specimens that are tested, we could observe a value of test statistics \bar{x} that falls into the **critical region(rejection region)**.
- We would then reject the null hypothesis H_0 in favor of the alternate H_1 , in fact, H_0 is really true
- This type of wrong conclusion is called a type I error

Type I Error

- Rejecting the null hypothesis H_0 when it is true is defined as a type I error

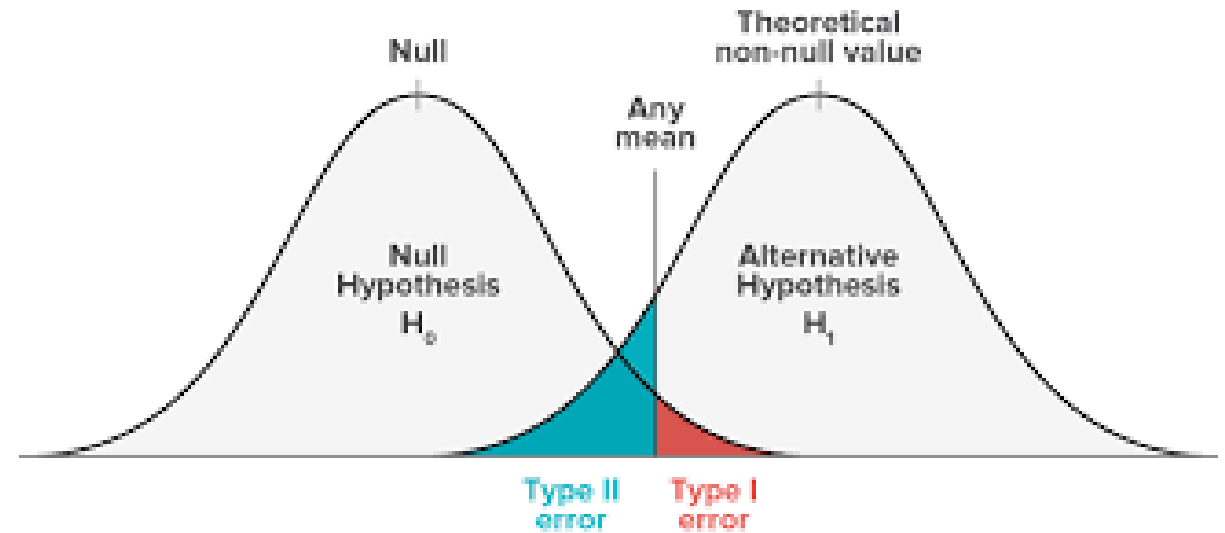


Type II Error

- Now suppose the true mean burning rate is different from 50 centimeters per second, yet the sample —
mean \bar{x} falls in the **acceptance region**
- In this case we would fail to reject H_0 when it is false
- This type of wrong conclusion is called a type II error

Type II Error

- Failing to reject the null hypothesis when it is false is defined as a type II error



Type 1 and Type II Errors

	H_0 is correct	H_0 is incorrect
H_0 is accepted	correct decision	Type II error (β) Incorrect acceptance
H_0 is rejected	Type I error (α) Incorrect rejection	correct decision

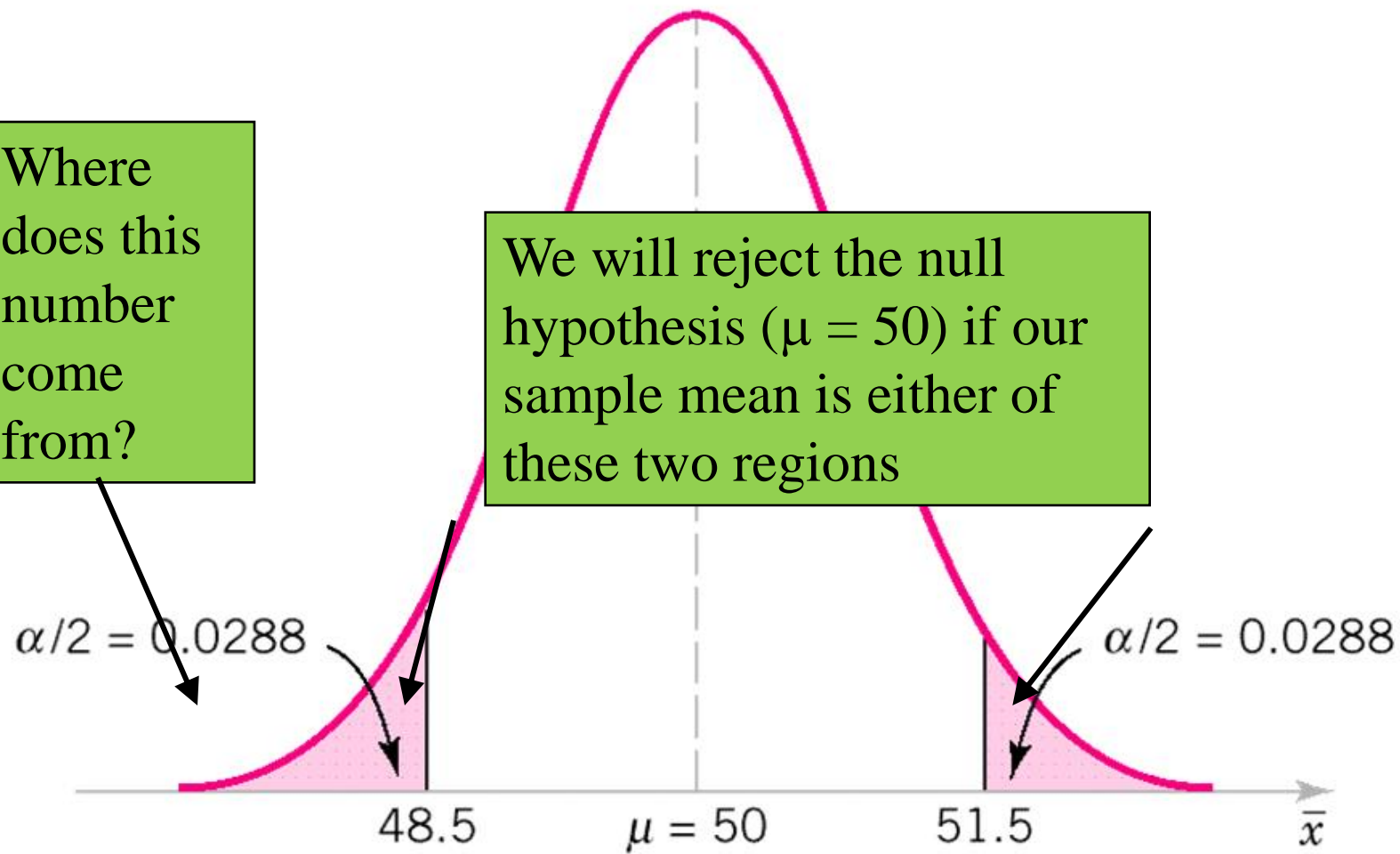
Type I error

- In the propellant burning rate example, a type I error will occur when either $\bar{x} > 51.5$ *or* $\bar{x} < 48.5$ when the true mean burning rate is $\mu = 50$ centimeters per second
- Suppose the standard deviation of burning rate is $\sigma = 2.5$ centimeters per second and $n = 10$
- Probability distribution $\mu = 50$, standard error = 0.79.
- Type I error is

$$\alpha = P(\bar{x} < 48.5 \text{ when } \mu = 50) + P(\bar{x} > 51.5 \text{ when } \mu = 50)$$

Where does this number come from?

We will reject the null hypothesis ($\mu = 50$) if our sample mean is either of these two regions



The critical region for $H_0: \mu = 50$ versus $H_1: \mu \neq 50$ and $n = 10$.

Defining function for calculating alpha value

```
In [6]: def z_value(x,mu,SEM):  
        z = (x - mu)/SEM  
        if(z < 0):  
            alfa = stats.norm.cdf(z)  
        else:  
            alfa = 1 - stats.norm.cdf(z)  
        print (alfa)
```

calculating alpha for different values of x,mu, and SEM

```
In [8]: x =48.5  
        mu = 50  
        SEM = 0.79
```

```
In [9]: z_value(x,mu,SEM)  
  
0.02879971774715278
```

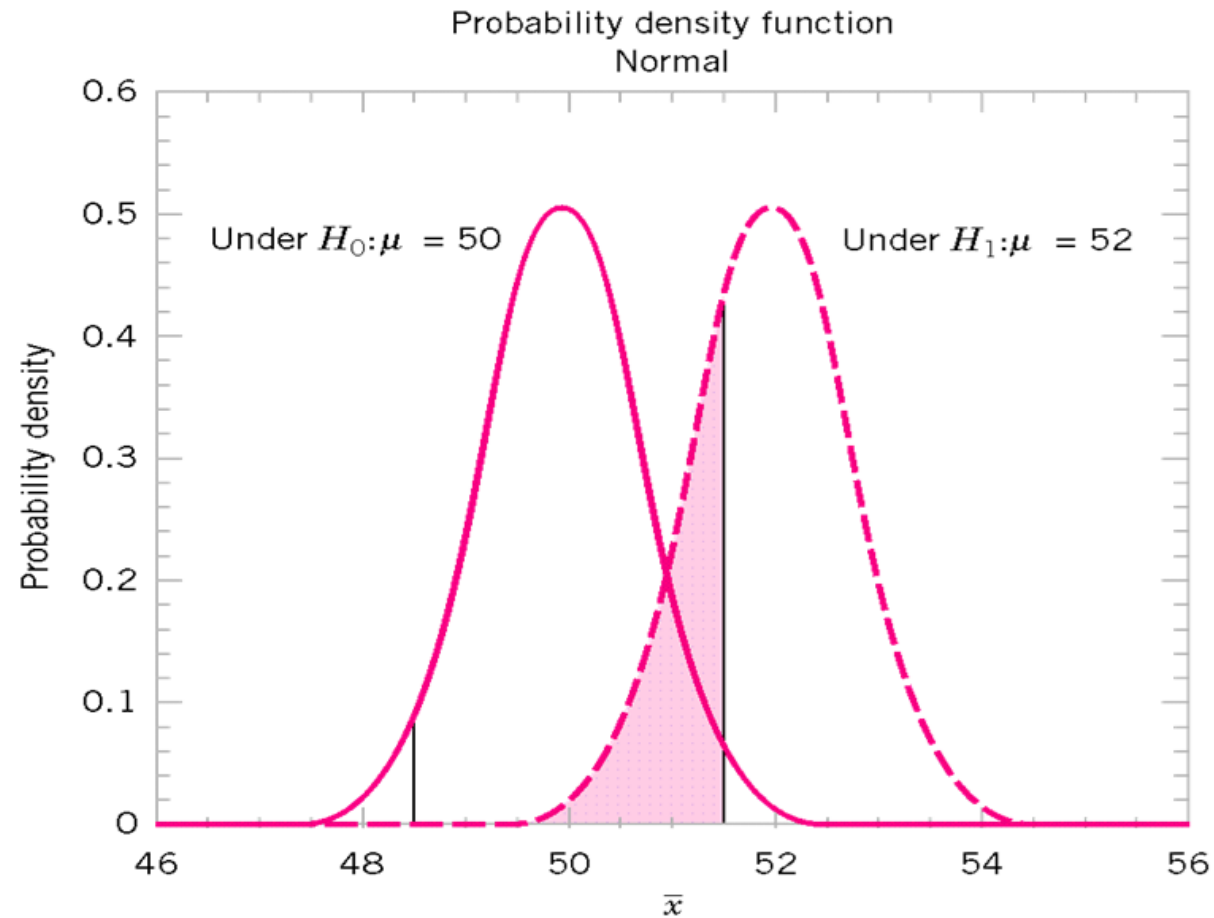
Type I error

- Type I error = 0.057434
- This implies that 5.7 % of all random samples would lead to rejection of the hypothesis $H_0: \mu=50$ centimeters per second.
- We can reduce the type I error by widening the acceptance region. If we make **critical value 48** and **52**, the value of alpha is 0.0114 (adding 0.0057 and 0.0057).
- Change sample size to 16 then alpha is 0.0164.

```
In [40]: z_value(48,mu,SEM)
0.005676434117424844
```

```
In [41]: z_value(52,mu,SEM)
0.0056764341174248
```

TYPE II ERROR



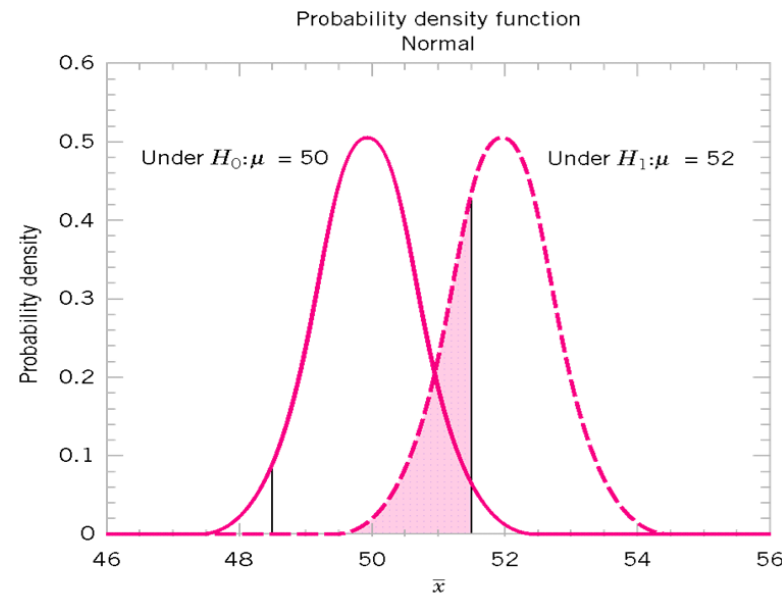
The probability of type II error when $\mu = 52$ and $n = 10$.

The pink area is
the probability
of a Type II error
if the actual mean
is 52.

Type II Error

- Type II error will be committed if the sample mean \bar{x} falls between 48.5 and 51.5 (critical region boundaries) when $\mu = 52$.
$$\beta = P(48.5 \leq \bar{x} \leq 51.5 \text{ when } \mu = 52)$$

- 0.2643
- When $\mu = 50.5$
- 0.8923



The probability of type II error when $\mu = 52$ and $n = 10$.

```
In [4]: beta = stats.norm.cdf((51.5-52)/0.79) #
```

```
In [5]: beta
```

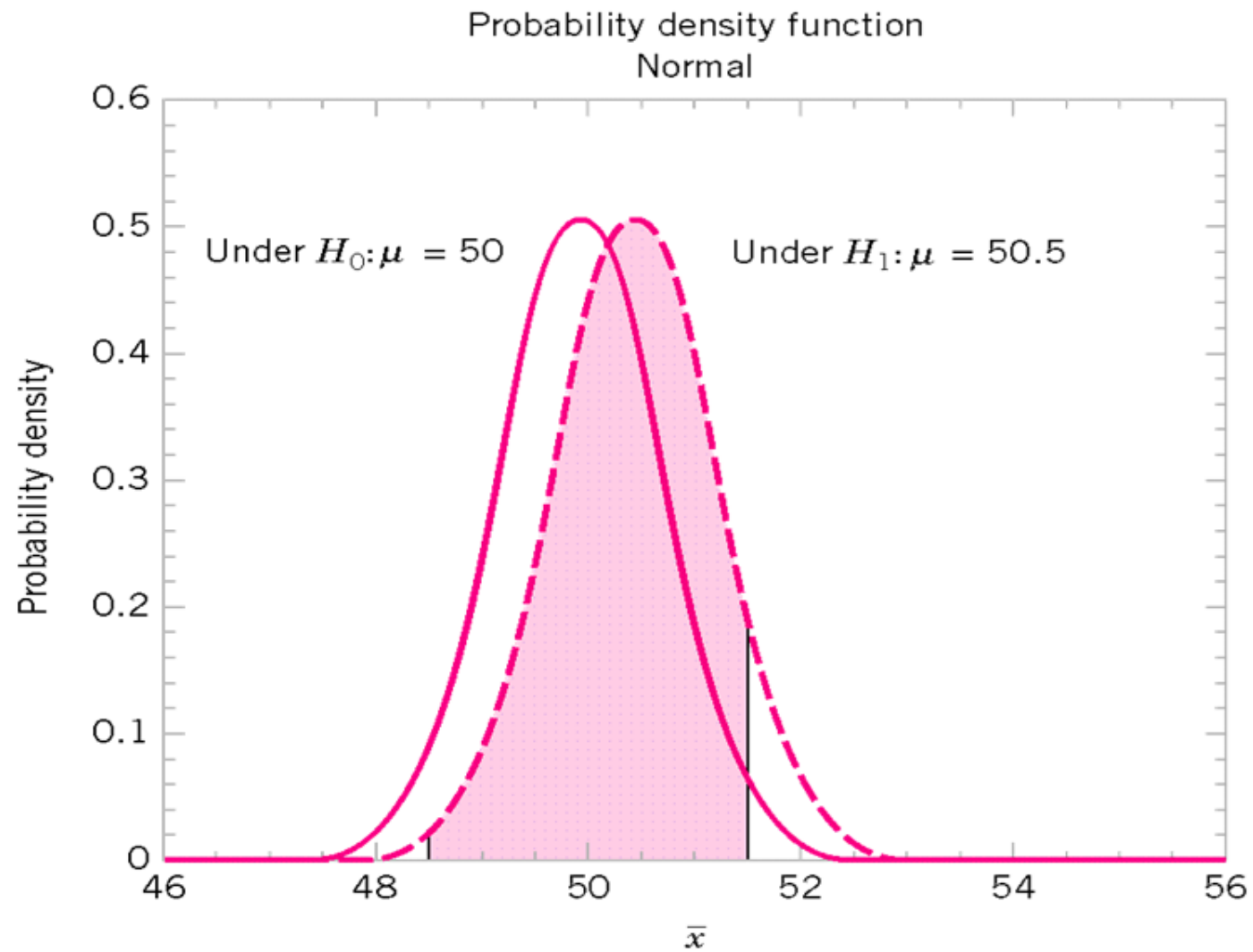
```
Out[5]: 0.26339575390741593
```

```
In [8]: beta = stats.norm.cdf((51.5-50.5)/0.79) #
```

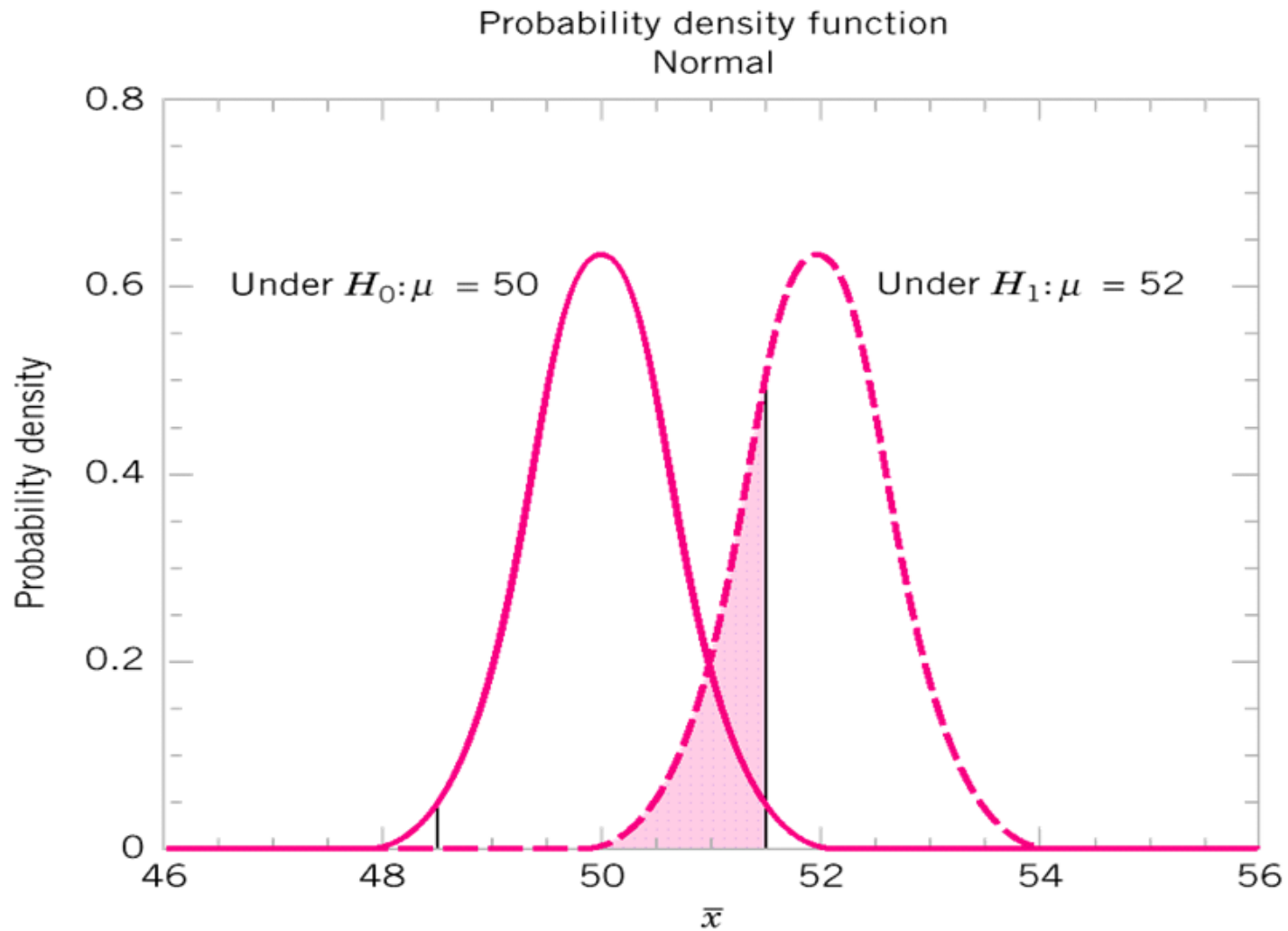
```
In [9]: beta
```

```
Out[9]: 0.8972117321157791
```



The probability of type II error when $\mu = 50.5$ and $n = 10$.



The probability of type II error when $\mu = 52$ and $n = 16$.

Computing the probability of a type II error may be the most difficult concept

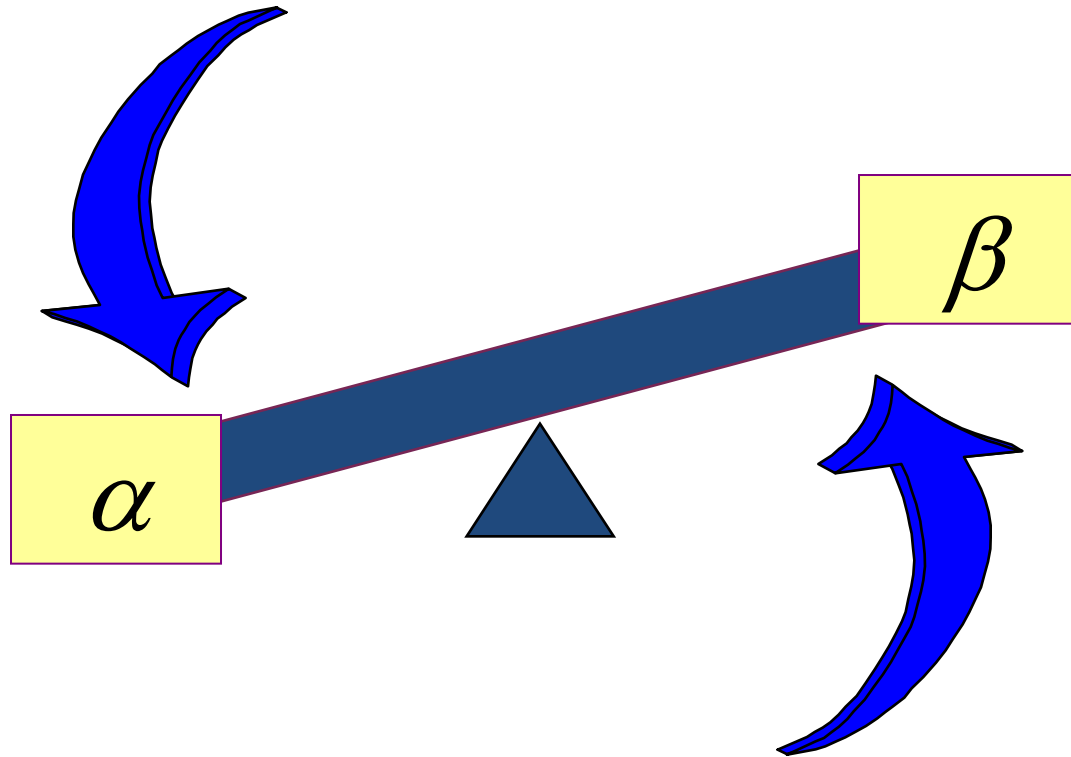
acceptance region	sample size	α	β at $\mu = 52$	β at $\mu = 50.5$
$48.5 < \bar{x} < 51.5$	10	0.0576	0.2643	0.8923
$48 < \bar{x} < 52$	10	0.0114	0.5000	0.9705
$48.5 < \bar{x} < 51.5$	16	0.0164	0.2119	0.9445
$48 < \bar{x} < 52$	16	0.0014	0.5000	0.9918

For constant n , increasing the acceptance region (hence decreasing α) increases β .

Increasing n , can decrease both types of errors.

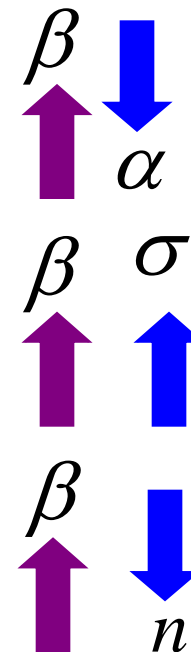
Type I & II Errors Have an Inverse Relationship

If you reduce the probability of one error, the other one increases so that everything else is unchanged.



Factors Affecting Type II Error

- True value of population parameter
 - β Increases when the difference between hypothesized parameter and its true value decrease
- Significance level α
 - Increases when β decreases
- Population standard deviation σ
 - Increases when β increases
- Sample size
 - β Increases when n decreases



How to Choose between Type I and Type II Errors

- Choice depends on the cost of the errors
- Choose smaller Type I Error when the cost of rejecting the maintained hypothesis is high
 - A criminal trial: convicting an innocent person
- Choose larger Type I Error when you have an interest in changing the status quo

Calculating the probability of Type II Error

$$H_0: \mu = 8.3$$

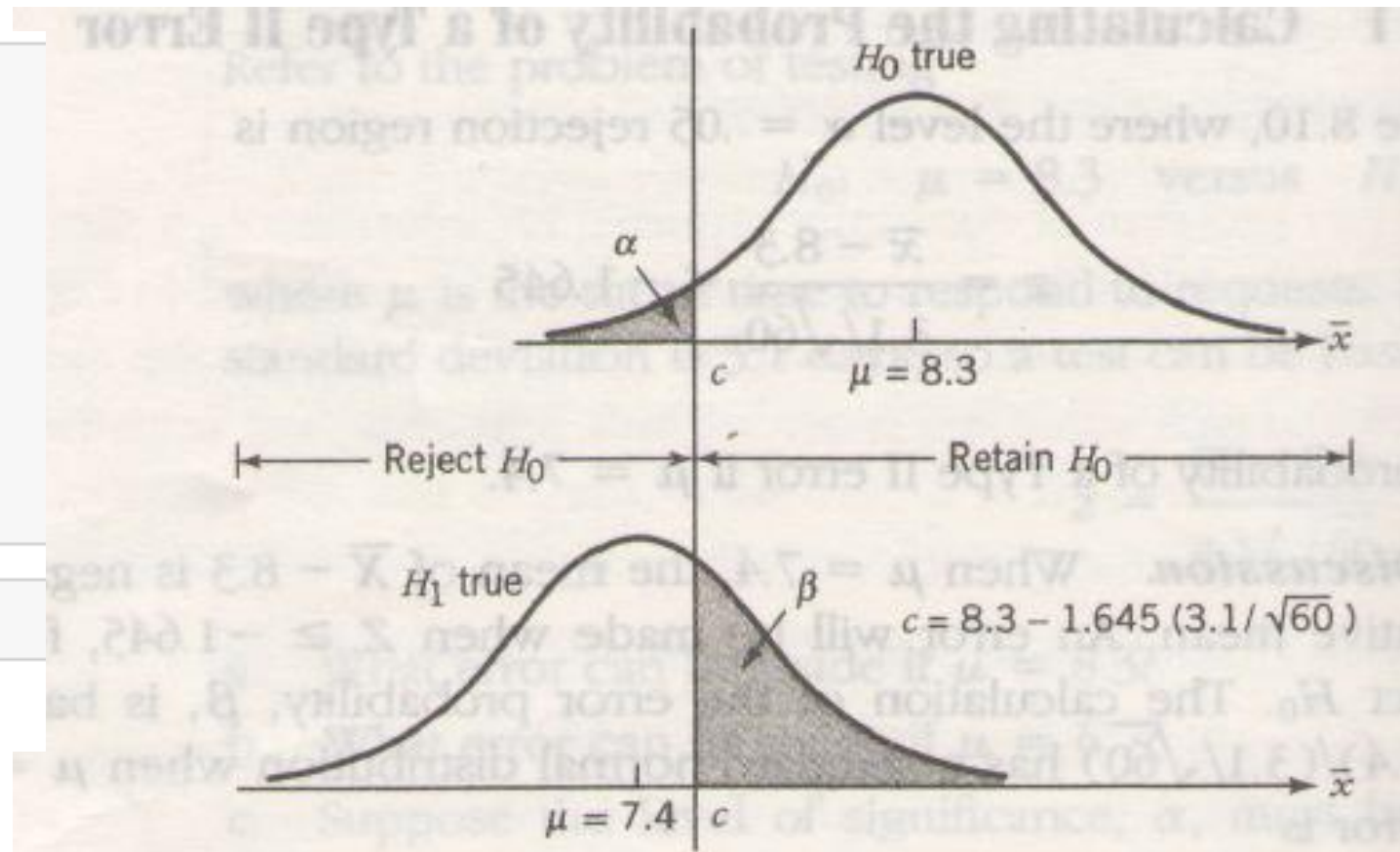
$$H_1: \mu < 8.3$$

Determine the probability of Type II error if $\mu = 7.4$ at 5% significance level. $\sigma = 3.1$ and $n = 60$.

Solution:

```
In [48]: def type_2(mu1,mu2,sigma,n,alfa):  
         z = stats.norm.ppf(alfa)  
         xbar = (mu1)+(z*sigma/np.sqrt(n))  
         z2 = (xbar - mu2)/(sigma/np.sqrt(n))  
         if(mu1 > mu2):  
             beta= 1-stats.norm.cdf(z2)  
         else:  
             beta = stats.norm.cdf(z2)  
         print (beta)
```

```
In [49]: type_2(8.3,7.4,3.1,60,0.05)  
0.27292999450730004
```



An error will be made when $Z \geq -1.645$, for that will fail to reject H_0 .

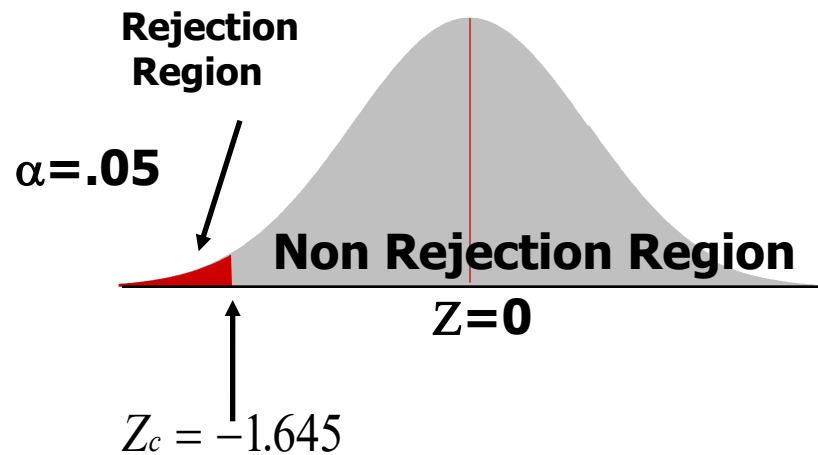
$$\beta = 0.2729$$

Solving for Type II Errors: Example

$$H_0: \mu = 12$$

$$H_a: \mu < 12$$

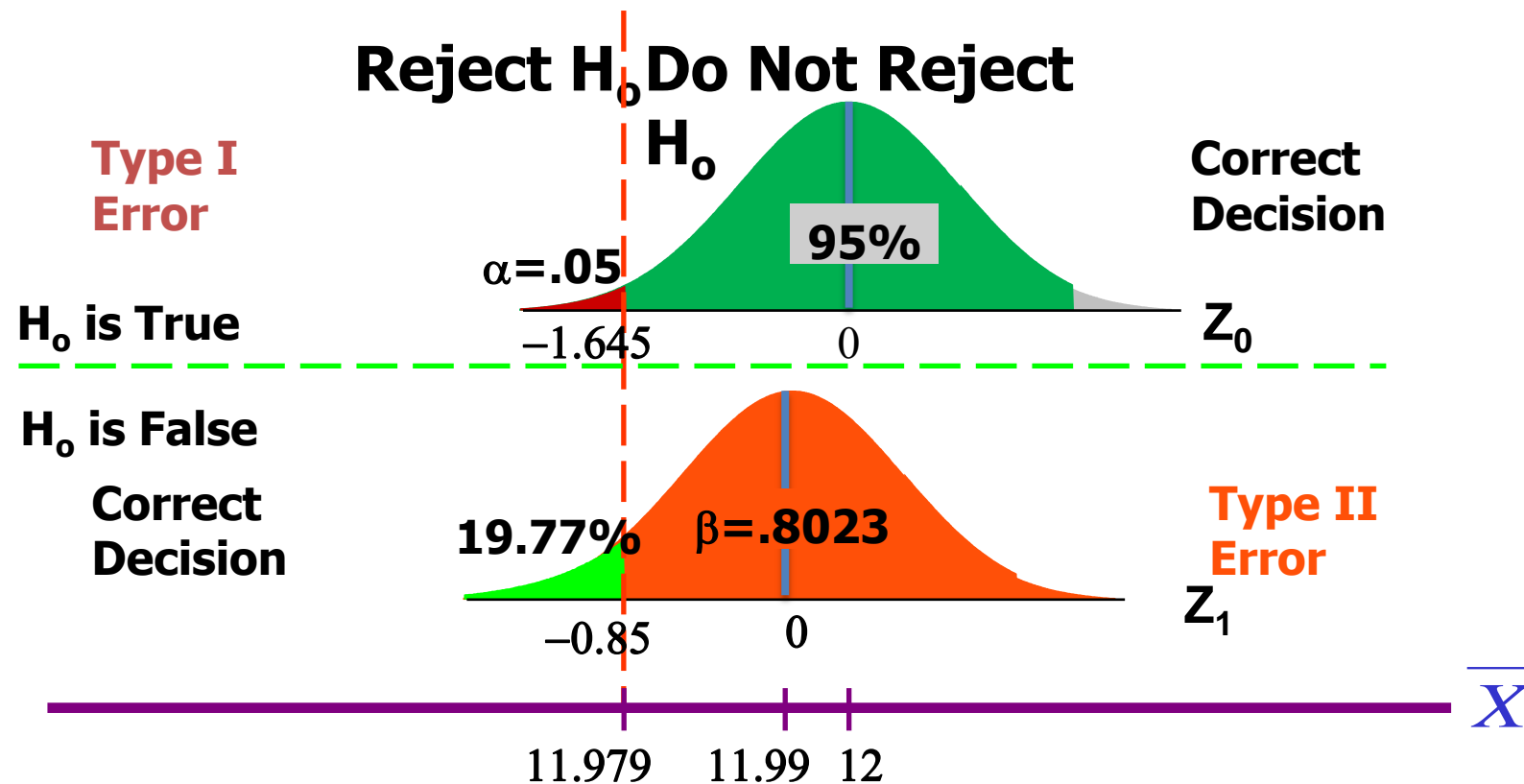
$$\begin{aligned}\bar{X}_c &= \mu + Z_c \frac{\sigma}{\sqrt{n}} \\ &= 12 + (-1.645) \frac{0.10}{\sqrt{60}} \\ &= 11.979\end{aligned}$$



If $\bar{X} < 11.979$, reject H_0 .

If $\bar{X} \geq 11.979$, do not reject H_0 .

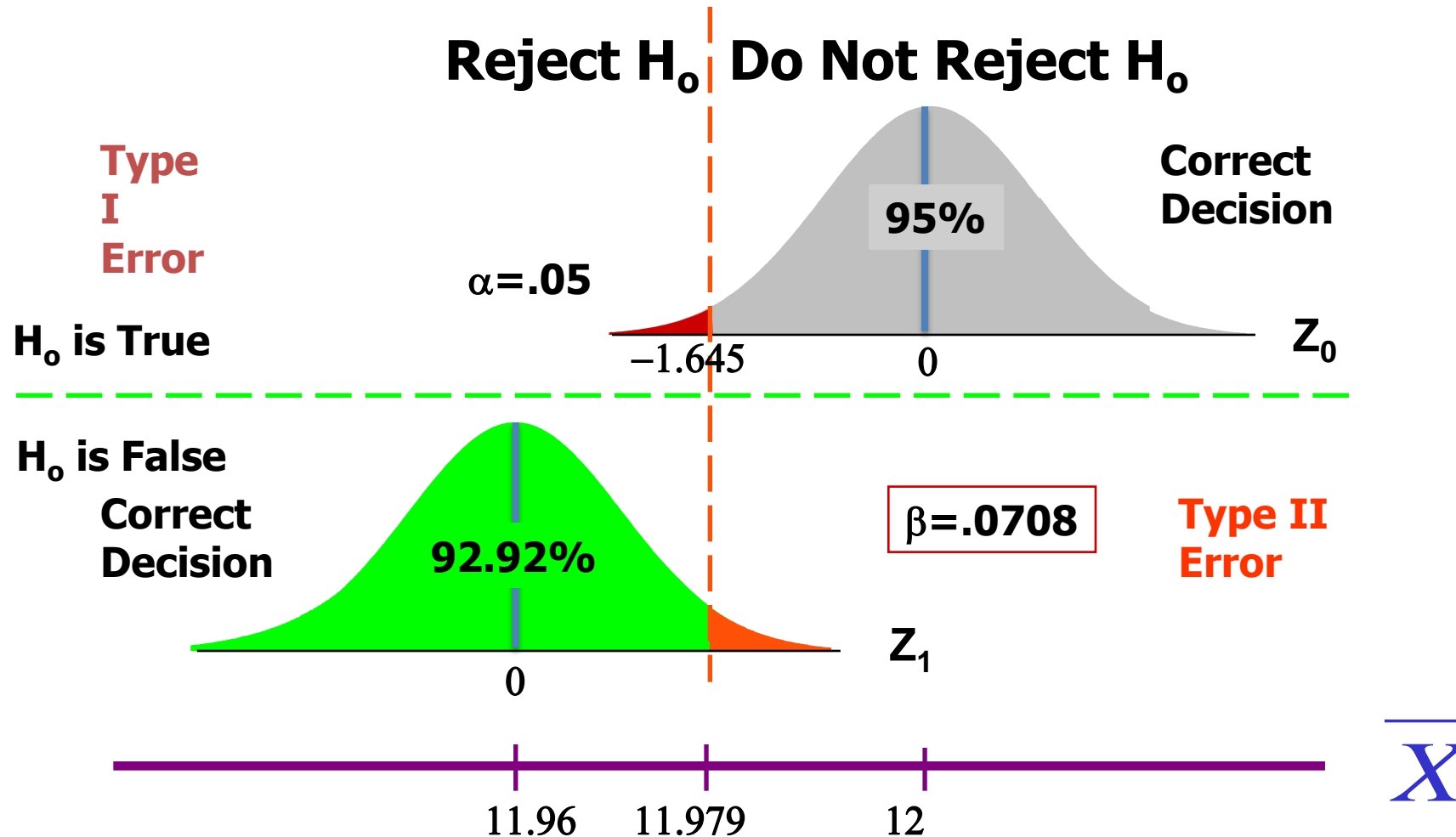
Type II Error for Example with $\mu = 11.99$ Kg



```
In [50]: type_2(12,11.99,0.1,60,0.05)
```

```
0.8079200023112734
```

Type II Error for Demonstration with $\mu=11.96$ Kg



```
In [51]: type_2(12,11.96,0.1,60,0.05)
```

```
0.07303790512847008
```

Hypothesis Testing and Decision Making

- We have illustrated hypothesis testing applications referred to as significance tests
- In the tests, we compared the p -value to a controlled probability of a Type I error, α , which is called the level of significance for the test
- With a significance test, we control the probability of making the Type I error, but not the Type II error
- We recommended the conclusion “do not reject H_0 ” rather than “accept H_0 ” because the latter puts us at risk of making a Type II error

Hypothesis Testing and Decision Making

- With the conclusion “do not reject H_0 ”, the statistical evidence is considered inconclusive
- Usually this is an indication to postpone a decision until further research and testing is undertaken
- In many decision-making situations the decision maker may want, and in some cases may be forced, to take action with both the conclusion “do not reject H_0 ” and the conclusion “reject H_0 .”
- In such situations, it is recommended that the hypothesis-testing procedure be extended to include consideration of making a Type II error

Power of a test

- The mean response time for a random sample of 40 food-order is 13.25 minutes
- The population standard deviation is believed to be 3.2 minutes.
- The restaurant owner wants to perform a hypothesis test, with $\alpha = 0.05$ level of significance, to determine whether the service goal of 12 minutes or less is being achieved.



Calculating the Probability of a Type II Error

Hypotheses are: $H_0: \mu \leq 12$ and $H_a: \mu > 12$

Rejection rule is: Reject H_0 if $z \geq 1.645$

Value of the sample mean that identifies the rejection region:

$$z = \frac{\bar{x} - 12}{3.2/\sqrt{40}} \geq 1.645$$

$$\bar{x} \geq 12 + 1.645 \left(\frac{3.2}{\sqrt{40}} \right) = 12.8323$$

We will accept H_0 when $x \leq 12.8323$

Calculating the Probability of a Type II Error

Probabilities that the sample mean will be in the acceptance region:

Values of μ	$z = \frac{12.8323 - \mu}{3.2/\sqrt{40}}$	β	$1-\beta$
14.0	-2.31	.0104	.9896
13.6	-1.52	.0643	.9357
13.2	-0.73	.2327	.7673
12.8323	0.00	.5000	.5000
12.8	0.06	.5239	.4761
12.4	0.85	.8023	.1977
12.0001	1.645	.9500	.0500

```
In [20]: type_2(14,12,3.2,40,0.05)  
0.010499750448532241
```

```
In [21]: type_2(13.6,12,3.2,40,0.05)  
0.06457982995225997
```

```
In [23]: type_2(13.2,12,3.2,40,0.05)  
0.2336575101104159
```

```
In [22]: type_2(12.8323,12,3.2,40,0.05)  
0.49995065746353273
```

```
In [27]: type_2(12.8,12,3.2,40,0.05)  
0.5254013387545549
```

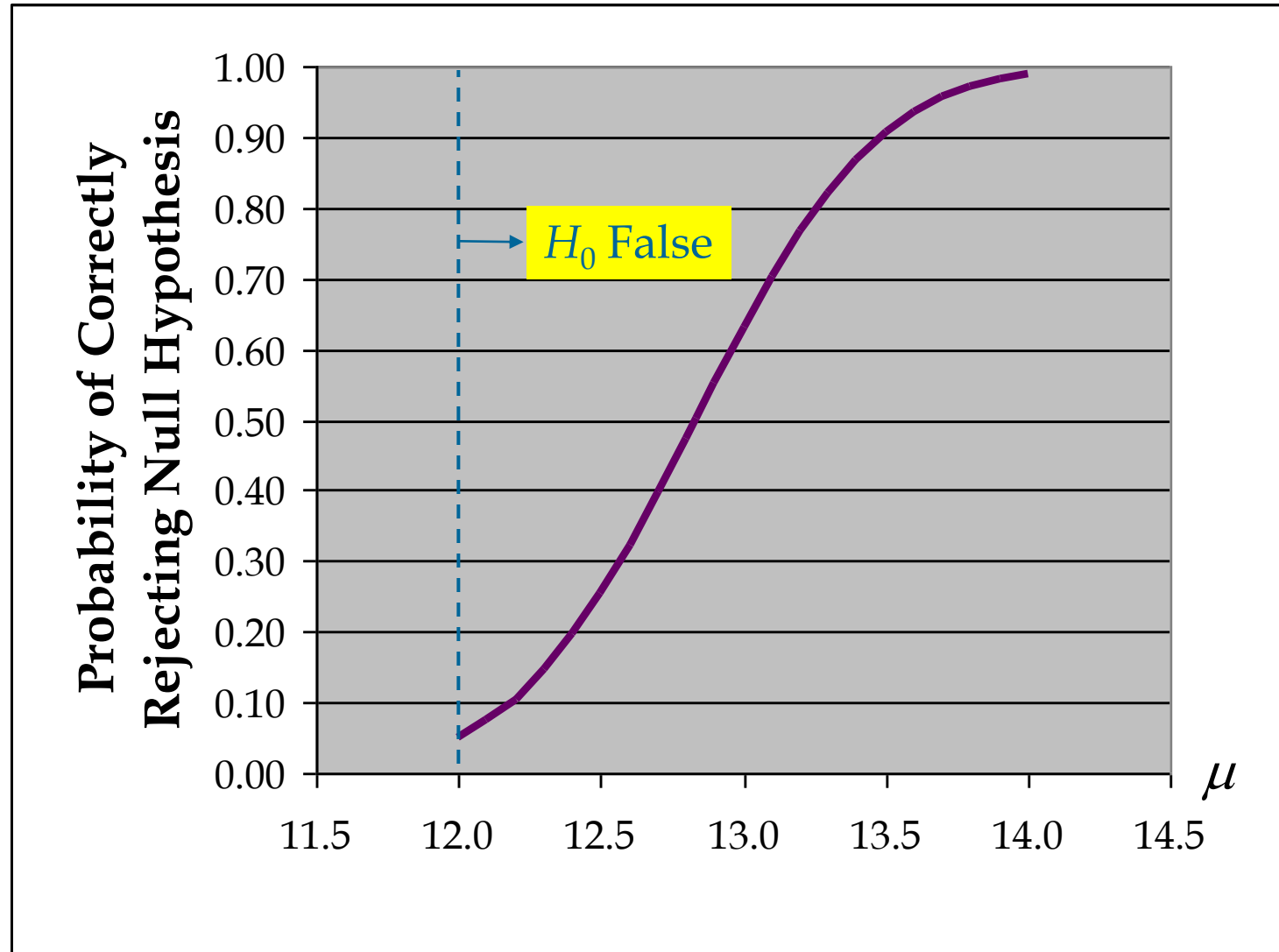
```
In [24]: type_2(12.4,12,3.2,40,0.05)  
0.8035262335707292
```

```
In [26]: type_2(12.0001,12,3.2,40,0.05)  
0.9499796127157129
```

Power of the Test

- The probability of correctly rejecting H_0 when it is false is called the power of the test.
- For any particular value of m , the power is $1 - b$.
- We can show graphically the power associated with each value of μ ; such a graph is called a power curve.

Power Curve



Thank You

