

# Statistical Hypothesis Testing – recap –

Thursday 1<sup>st</sup> December, 2016

# Experimenting

perform an experiment - to discover something

setting	experiment	discovery
alarm system	check sensors	intruder
new treatment for illness	test treatment on people	treatment works
hacker attack detection	test network traffic	there is an attack
software testing	run tests	bug

# Alarm System

discovery = intruder

	test output = negative	test output = positive
no intruder	true negative	false positive
intruder	false negative	true positive

# New Treatment for Illness

discovery = treatment works

	test output = negative	test output = positive
treatment doesn't work	true negative	false positive
treatment works	false negative	true positive

# Hacker Attack Detection

discovery = hacker attack

	test output = negative	test output = positive
no attack	true negative	false positive
attack	false negative	true positive

# Software Testing

discovery = bug

	test output = negative	test output = positive
no bug	true negative	false positive
bug	false negative	true positive

If we know what a discovery looks like

## Alarm system

sensor readings  $X$

no intruder	$0 \leq X \leq 4$
intruder	$5 \leq X \leq 8$

decision rule  $X \stackrel{?}{>} 4$

no	no intruder
yes	intruder

If we know what a discovery looks like

## New Treatment

number of tails  $T$

sick	$T > 1$
not sick	$T = 0$

decision rule (treatment works if)  $T \stackrel{?}{\leq} 1$

no	doesn't work
yes	work



If we know what a discovery looks like

## Hacker Attack Detection

average upload, idle computer  $U$

no attack	$U < 30 [Kb/s]$
attack	$U > 50 [Kb/s]$

decision rule (attack if)  $U \stackrel{?}{\geq} 30$

no	no attack
yes	attack

If we know what a discovery looks like

## Software Testing

number of failed tests  $N_f$

no bug	$N_f = 0$
bug	$N_f > 0$

decision rule (there is a bug if)  $N_f \stackrel{?}{>} 0$

no	no bug
yes	bug

If we don't know what a discovery looks like

## Alarm system

sensor readings  $X$

no intruder	$0 \leq X \leq 4$
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decision rule  $X \stackrel{?}{>} 4$

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yes	intruder

If we don't know what a discovery looks like

## New Treatment

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yes	attack

If we don't know what a discovery looks like

## Software Testing

number of failed tests  $N_f$

no bug	$N_f = 0$
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decision rule (there is a bug if)  $N_f \stackrel{?}{>} 0$

no	no bug
yes	bug

# Experimenting with Distributions

- ▶  $N_1 = \mathcal{N}(\mu, \sigma^2)$ ,  $\mu_0$  constant
- ▶  $Y$  fair coin

setting	experiment	discovery
is $X$ not sampled from $N_1$ ?	?	$X$ not sampled from $N_1$
is $Y$ unfair?	?	$Y$ is unfair
is $\mu > \mu_0$ ?	?	$\mu > \mu_0$
$X$ and $Y$ are dependent?	?	$X$ and $Y$ are dependent

Is  $X$  not sampled from  $N_1$

discovery =  $X$  not sampled from  $N_1$

	test output = negative	test output = positive
$X \sim N_1$	true negative	false positive
$X \not\sim N_1$	false negative	true positive



# Is $Y$ unfair?

discovery =  $Y$  is unfair

	test output = negative	test output = positive
$Y$ is fair	true negative	false positive
$Y$ is unfair	false negative	true positive

Is  $\mu > \mu_0$ ?

discovery =  $\mu > \mu_0$

	test output = negative	test output = positive
$\mu \leq \mu_0$	true negative	false positive
$\mu > \mu_0$	false negative	true positive

# Are $X$ and $Y$ dependent?

discovery =  $X$  and  $Y$  are dependent

	test output = negative	test output = positive
independent	true negative	false positive
dependent	false negative	true positive

If we know what a discovery looks like

**Is  $X$  not sampled from  $N_1$ ?**

$X$  values

sampld from $N_1$	$X \sim N_1$
sampld from $N_2 = \mathcal{N}(\mu_2, \sigma^2)$	$X \sim N_2$

decision rule (MLE)  $|X - \mu|^2 \stackrel{?}{>} |X - \mu_2|^2$

no	sampld from $N_1$
yes	not sampld from $N_1$

If we know what a discovery looks like

**Is  $Y$  unfair?**

coin tosses  $Y_1, \dots, Y_n$ ,  $[\sum_{i=1}^n Y_i] = c$

fair	$P([\sum_{i=1}^n Y_i] = c) = \binom{n}{c} 0.5^n$
$P(Y = 1) = 0.8$	$P([\sum_{i=1}^n Y_i] = c) = \binom{n}{c} 0.8^c 0.2^{n-c}$

decision rule (MLE)  $\binom{n}{c} 0.8^c 0.2^{n-c} > \binom{n}{c} 0.5^n$

no	fair
yes	unfair

If we know what a discovery looks like

**Is  $\mu > \mu_0$ ?**

sample mean  $\bar{X} = \sum_{i=1}^n X_i/n$

$\mu = \mu_0$	$\bar{X} \sim \mathcal{N}(\mu_0, \sigma^2/n)$
$\mu = \mu_1 > \mu_0$	$\bar{X} \sim \mathcal{N}(\mu_1, \sigma^2/n)$

decision rule (MLE)  $P(\bar{X} \mid \mu_1) > P(\bar{X} \mid \mu_0)$

no	$\mu = \mu_0$
yes	$\mu > \mu_0$

# Logical Acrobatics

If we **know** what a discovery looks like

- ▶ we know the "normal" explanation hypothesis
- ▶ we know the alternative explanation hypothesis
- ▶ decision rule:

probability of  
results given the  
"normal"  
explanation

<

probability of  
results given the  
alternative  
explanation

inaccurate

# Logical Acrobatics

If we **know** what a discovery looks like

- ▶ we know the "normal" explanation hypothesis
- ▶ we know the alternative explanation hypothesis
- ▶ decision rule:

improbability of  
results given the  
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explanation

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improbability of  
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inaccurate



# Logical Acrobatics

If we **don't know** what a discovery looks like

- ▶ we know the "normal" explanation hypothesis
- ▶ we don't know the alternative explanation hypothesis
- ▶ decision rule:  
improbability of  
results given the  
"normal"  
explanation  $>$  threshold  $th'$

inaccurate

# Logical Acrobatics

If we **don't know** what a discovery looks like

- ▶ we know the "normal" explanation hypothesis
- ▶ we don't know the alternative explanation hypothesis
- ▶ decision rule:

probability of  
results given the  
"normal"  
explanation  $<$  threshold  $th$

inaccurate

# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$

reject "normal"  
explanation

but

"normal"  
explanation is true

# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$

reject "normal"  
explanation

but

"normal"  
explanation is true

- ▶ alarm system

reject -  
"no intruder"

but

there is no intruder

# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$

reject "normal"  
explanation

but

"normal"  
explanation is true

- ▶ testing new treatment

reject -  
"treatment doesn't work"

but

treatment  
doesn't work

# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$   
reject "normal"  
explanation      but      "normal"  
explanation is true
- ▶ hacker attack detection  
reject -  
"no attack"      but      there is no attack

# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$   
reject "normal"  
explanation      but      "normal"  
explanation is true
- ▶ software testing  
reject -  
"no bug"      but      there is no bug

# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$

reject "normal"  
explanation

but

"normal"  
explanation is true

- ▶ Is  $X$  not sampled from  $N_1$ ?

reject -  
 $X$  sampled from  $N_1$

but

$X$  is sampled from  
 $N_1$

double negative - board



# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$

reject "normal"  
explanation

but

"normal"  
explanation is true

- ▶ is  $Y$  unfair?

reject -  
 $Y$  is fair

but

$Y$  is fair

double negative

# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$

reject "normal"  
explanation

but

"normal"  
explanation is true

- ▶ Is  $\mu > \mu_0$ ?

reject -  
 $\mu \leq \mu_0$

but

$\mu \leq \mu_0$

# Probability of False Positive

a false positive

- ▶ "normal" explanation -  $H_0$

reject "normal"  
explanation

but

"normal"  
explanation is true

- ▶  $X$  and  $Y$  are dependent?

reject -  
 $X$  and  $Y$  are  
independent

but

$X$  and  $Y$  are  
independent

# Rejected Results

- ▶ we reject "normal" explanation  $H_0$  for some results
  - testing treatment - 95% of patients cured
  - software testing - 8/10 tests fail
  - testing a coin - 9/10 heads
- ▶ we do not reject for others
  - testing treatment - 0.5% of patients cured
  - software testing - no tests fail
  - testing a coin - 6/10 heads

# Rejected Results

- ▶ we reject "normal" explanation  $H_0$  for some results
- ▶ we do not reject for others
- ▶ results we reject
  - are possible given "normal" explanation  $H_0$
  - just unlikely

# Rejected Results

- ▶ we reject "normal" explanation  $H_0$  for some results
- ▶ we do not reject for others
- ▶ results we reject
  - are possible given "normal" explanation  $H_0$
  - just unlikely

probability of results we reject  
given the "normal" explanation  $H_0$  = probability of  
false positives

# Rejected Results

- ▶ if we reject for
  - testing treatment - 95% of patients cured
  - software testing - 8/10 tests fail
  - testing a coin - 9/10 heads
- ▶ we will also reject for
  - testing treatment - 99% of patients cured
  - software testing - 9/10 tests fail
  - testing a coin - 10/10 heads

# Rejected Results

- ▶ if we consider
  - testing a coin - 9/10 heads
- ▶ unlikely enough to reject "normal" explanation  $H_0$
- ▶ we would also consider
  - 10/10 heads
  - 1/10 heads
  - 0/10 heads
- ▶ unlikely enough to reject



# P-Value

- ▶ this value for 9/10 heads

$$\begin{aligned} f(9/10) = & P(9/10 \mid H_0) + P(10/10 \mid H_0) \\ & + P(1/10 \mid H_0) + P(0/10 \mid H_0) \end{aligned}$$

- ▶ is the P-Value for 9/10 heads

# P-Value

- ▶ result  $X$
- ▶ all results at least as unlikely (improbable)

$$R(X) = \{Z : P(Z | H_0) \leq P(X | H_0)\}$$

# P-Value

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$$R(X) = \{Z : P(Z | H_0) \leq P(X | H_0)\}$$

- ▶ P-Value

$$\begin{aligned} Pval(X) &= \sum_{Z \in R(X)} P(Z | H_0) \\ &= P(R(X) | H_0) \end{aligned}$$

# How do we choose a threshold?

we are willing to allow 5% false positives

- ▶ if  $Pval(X) < th = 0.05$ 
  - reject "normal" explanation  $H_0$  for result  $X$
- ▶ else
  - do no reject

Does this guarantee a false positive rate?

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challenge

- ▶  $X_{th}$  the most likely result we consider unlikely  
multiple possible
  - example: 9/10 heads

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  - 9/10 heads, 10/10 heads, 1/10 heads, 0/10 heads

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- ▶ probability of false positive
  - probability of anything at least as unlikely as  $X_{th}$
  - $Pval(X_{th})$  P-Value of  $X_{th}$



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  - example: 9/10 heads
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  - 9/10 heads, 10/10 heads, 1/10 heads, 0/10 heads
- ▶ probability of false positive
  - probability of anything at least as unlikely as  $X_{th}$
  - $Pval(X_{th})$  P-Value of  $X_{th}$

$$Pval(X_{th}) = P(\text{reject } H_0 \mid H_0) < th$$

# How do we choose the null hypothesis $H_0$

- ▶ usually stands for "normality"
  - no surprise
  - no change
  - treatment doesn't work
  - no intruder
  - ...

# How do we choose the null hypothesis $H_0$

- ▶ usually stands for "normality"
  - no surprise
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  - treatment doesn't work
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  - ...
- ▶ always has a computable distribution
  - we can compute  $P(X | H_0)$

# Null hypothesis examples

setting	experiment	null hypothesis
alarm system	check sensors	
new treatment for illness	test treatment on people	
hacker attack detection	test network traffic	
software testing	run tests	

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# Null hypothesis examples

setting	experiment	null hypothesis
alarm system	check sensors	no intruder
new treatment for illness	test treatment on people	treatment doesn't work
hacker attack detection	test network traffic	no attack
software testing	run tests	

# Null hypothesis examples

setting	experiment	null hypothesis
alarm system	check sensors	no intruder
new treatment for illness	test treatment on people	treatment doesn't work
hacker attack detection	test network traffic	no attack
software testing	run tests	no bug



# Null hypothesis examples

setting	experiment	null hypothesis
is $X$ not sampled from $N_1$ ?	sample $X$	
is $Y$ unfair?	toss coin	
is $\mu > \mu_0$ ?	sample $X$ , calculate sample mean	
$X$ and $Y$ are dependent?	sample $(X, Y)$	

# Null hypothesis examples

setting	experiment	null hypothesis
is $X$ not sampled from $N_1$ ?	sample $X$	$X \sim N_1$
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is $X$ not sampled from $N_1$ ?	sample $X$	$X \sim N_1$
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is $\mu > \mu_0$ ?	sample $X$ , calculate sample mean	$\mu \leq \mu_0$ $\bar{X} \sim N(\mu_0, \sigma^2/n)$
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is $X$ not sampled from $N_1$ ?	sample $X$	$X \sim N_1$
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is $\mu > \mu_0$ ?	sample $X$ , calculate sample mean	$\mu \leq \mu_0$ $\bar{X} \sim N(\mu_0, \sigma^2/n)$
$X$ and $Y$ are dependent?	sample $(X, Y)$	$X$ and $Y$ are independent

# Test statistic

**test statistic**  $T$

what we compute P-Value for

$$Pval(X) = P(T(X) \mid H_0)$$

# Test statistic

## test statistic $T$

what we compute P-Value for

example - Z-score

setting	experiment	test statistic*
is $X$ not sampled from $N_1$ ?	sample $X$	
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# Test statistic

## test statistic $T$

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example - Z-score

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# Test statistic

## test statistic $T$

what we compute P-Value for

example - Z-score

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is $X$ not sampled from $N_1$ ?	sample $X$	$X$
is $Y$ unfair?	toss coin	$\sum_{i=1}^n Y_i$
is $\mu > \mu_0$ ?	sample $X$ , calculate sample mean	
$X$ and $Y$ are dependent?	sample $(X, Y)$	

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## test statistic $T$

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example - Z-score

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is $X$ not sampled from $N_1$ ?	sample $X$	$X$
is $Y$ unfair?	toss coin	$\sum_{i=1}^n Y_i$
is $\mu > \mu_0$ ?	sample $X$ , calculate sample mean	$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$
$X$ and $Y$ are dependent?	sample $(X, Y)$	

# Test statistic

## test statistic $T$

what we compute P-Value for

example - Z-score

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is $X$ not sampled from $N_1$ ?	sample $X$	$X$
is $Y$ unfair?	toss coin	$\sum_{i=1}^n Y_i$
is $\mu > \mu_0$ ?	sample $X$ , calculate sample mean	$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$
$X$ and $Y$ are dependent?	sample $(X, Y)$	$2 \log \frac{P(X,Y)}{P(X)P(Y)}^*$

# Questions?

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