

# Randomization/Permutation tests

ST551 Lecture 28

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# Contingency Tables

	$Y_i=0$	$Y_i=1$
$G_i=0$	a	b
$G_i=1$	c	d

Chi-square

Fisher's

$$OR \rightarrow \frac{ad}{bc}$$

$$\frac{bc}{ad}$$

Two proportions z-test

$$\hat{p}_y \rightarrow P(Y_i=1 | G_i=0)$$

$$\hat{p}_x \rightarrow P(Y_i=1 | G_i=1)$$

# Announcements

I haven't received any suggestions for the formula sheet... draft on class webpage

Homeworks:

- 40% of your grade
- Lowest (%) HW dropped
- Remaining 8 homeworks will be weighted equally (i.e. 5% each)
- I'll update canvas with this contribution after HW #8 graded

Friday: no lecture, I'll be in my office.



# Randomized experiments

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# Two common study designs

## 1. Random Sampling study

→ Population Inferences  
 $Y_i$ : iid from some population

- A population(s) is defined
- Units are **randomly sampled** from the population(s)
- Units are observed

## 2. Randomized Experiment

→ Causal Inference

- A group of units is selected
- Units are **randomly assigned** to different levels of a treatment variable
- Units are observed





# Random Sampling Model



# KEY COMPONENTS

## Population Distribution



## Sample

Draw a random sample from population:

$$Y_1, \dots, Y_n$$

## Statistic

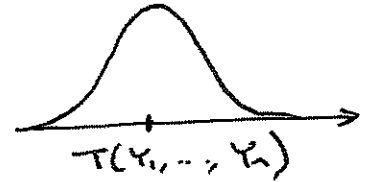
One number summary of the sample:

$$T(Y_1, \dots, Y_n)$$

eg.  $\bar{Y}$

## Sampling Distribution

Distribution of the statistic over all possible random samples.



SO FAR we have been working left to right  $\Rightarrow$

We know (or assume) the population dist.

+

Know sample size,  $n$  + Pick a statistic  $\equiv$   
Know the sampling mechanism

derive  
simulate  
approximate

Know (almost know) the sampling dist.

Relate properties to population properties

## INFERENCE

goes right to left  $\Leftarrow$

Make a statement about the population



We have one Sample:  
 $Y_1, \dots, Y_n$

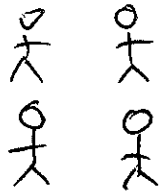
+ We calculate a statistic with the sample  $\equiv$

HOW? Use  $\Rightarrow$  relationship between pop. dist and sampling dist. to guide us



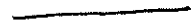
# Randomized Experiment Model

Experimental  
Units



Random assignment

Assign treatments  
to units at  
random

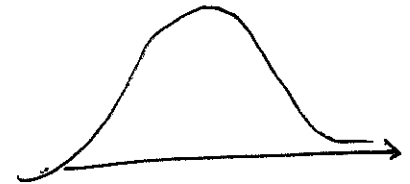


Statistic

One number  
summary of  
data

Randomization Dist

Distribution of  
statistic over  
all possible  
random assignments  
of units to  
treatments



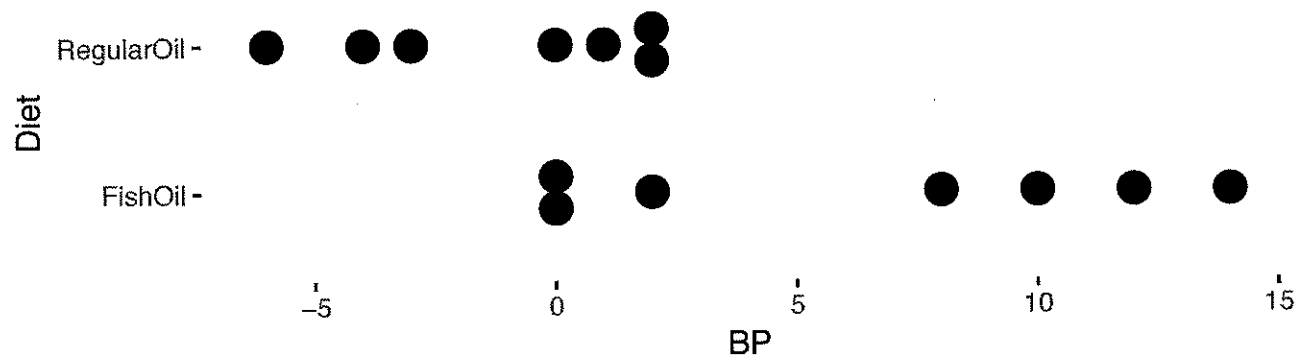


# Example

```
library(Sleuth3)
```

```
?ex0112
```

*Researchers used 7 red and 7 black playing cards to randomly assign 14 volunteer males with high blood pressure to one of two diets for four weeks: a fish oil diet and a standard oil diet. These data are the reductions in diastolic blood pressure.*







## Example

Did the fish oil decrease BP more than the Regular Oil?

*Sample means*

FishOil	RegularOil	FishOil - RegularOil
6.571	-1.143	7.714



# Randomization Distribution

The randomization distribution is the distribution of the statistic over all possible assignments of the treatments to the experimental units.

Just like the sampling distribution you can:

- derive it
- approximate it
- simulate it



# Simulating the Randomization Distribution

The usual null hypothesis in randomized experiments: no difference between treatments.

We observe pairs  $(Y_i, T_i)$  where  $Y_i$  is observed response, and  $T_i$  is the treatment applied (let's say  $T_i = 1$  or  $2$ ).

Often an additive model is assumed:

$$Y_i | (T_i = 2) = Y_i | (T_i = 1) + \delta$$

*treatment effect* ←

Under null  $\delta = 0$ , or if null is true, we observe  $Y_i = y_i$  regardless of the treatment unit  $i$  receives.

We only observe one of  $(Y_i, T_i = 1)$  or  $(Y_i, T_i = 2)$ , but if the null is true, we know what we would observe for person  $i$  under the other treatment, the same value.



## Example cont.

Null hypothesis: no difference between treatments

Alt. universe	BP	Subject	BP	Diet
Regular Oil	8	2	8	FishOil
		10	12	FishOil
		1	10	FishOil
		11	14	FishOil
			2	FishOil
			0	FishOil
			0	FishOil
			-6	RegularOil
			0	RegularOil
			1	RegularOil
			2	RegularOil
			-3	RegularOil
			-4	RegularOil
			2	RegularOil

Subject #2 randomly  
was assigned Fish Oil





# Example cont.

Null hypothesis: no difference between treatments

BP	Diet	random_1	random_2
8	FishOil	RegularOil	FishOil
12	FishOil	RegularOil	FishOil
10	FishOil	RegularOil	RegularOil
14	FishOil	RegularOil	FishOil
2	FishOil	RegularOil	RegularOil
0	FishOil	RegularOil	RegularOil
0	FishOil	FishOil	RegularOil
-6	RegularOil	RegularOil	FishOil
0	RegularOil	FishOil	RegularOil
1	RegularOil	FishOil	RegularOil
2	RegularOil	FishOil	FishOil
-3	RegularOil	FishOil	FishOil
-4	RegularOil	FishOil	RegularOil
2	RegularOil	FishOil	FishOil

$\binom{14}{7} \approx 3000$

$\overline{\text{Fish}} - \overline{\text{Regular}}$

7.714

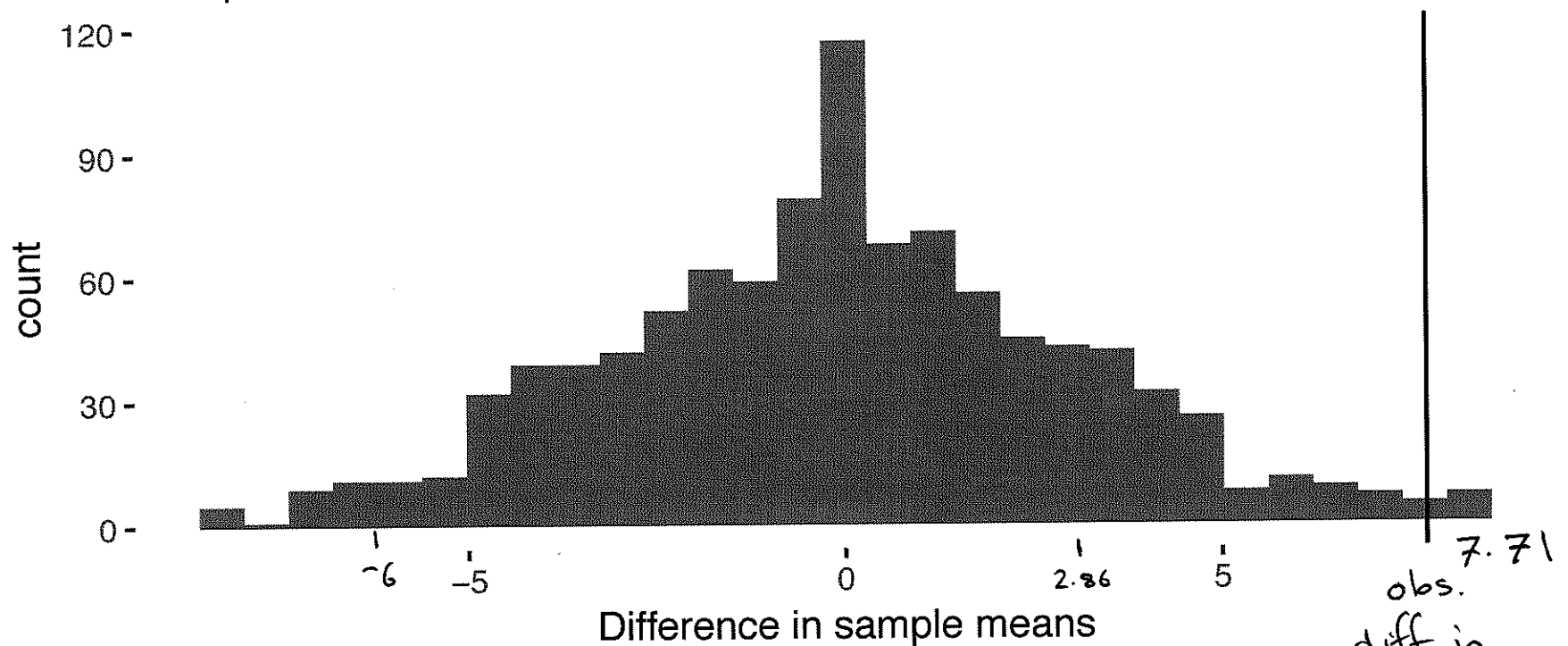
$\overline{\text{random}_1}$   
- 6.00

2.86



# Many permutations

Randomization distribution  
1000 permutations of treatment levels



prop. diff. in samples means  $\geq 7.71$

## [1] 0.007

one-sided p-value randomization test  
test stat = diff in sample means



# Randomization test

1. Pick a test statistic
2. Simulate the randomization distribution of the test statistic under all (or many) different assignments of the treatments  
Repeat many times:
  - 2.1 Permute treatment labels over observed values
  - 2.2 Recalculate test statistic
3. Compare the observed test statistic to the randomization distribution

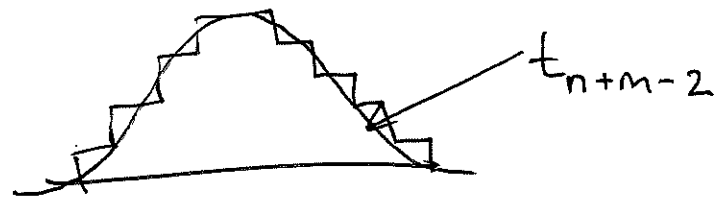
*p-value*  
*critical value*

It turns out :

If you use t-statistic as  
your test statistic  $\frac{\bar{Y} - \bar{X}}{\sqrt{s_p \left( \frac{1}{n} + \frac{1}{m} \right)}}$

If  $n$  is "large" and  $Y$  isn't too  
discrete,

the randomization dist is very  
close to a t-distribution



equivalent to a t-test.

## Randomization test: Comments

Exact? Consistent? Depends on the test statistic.

E.g. the test statistic 'difference in sample medians' isn't an exact test for equality of population medians unless we add an *additive effect* assumption.

Why? Reference distribution is calculated under the assumption that the values from the two groups are exchangeable.

Sometimes used with random sampling studies (often referred to as a permutation test). Pretends *population membership is like a random assignment*.





# The bigger picture

One sample t-test

$$Y_i \quad i=1, \dots, n$$

is a special case of regression

t-test on the coefficient  $\beta_0$  is the regression

$$Y_i = \beta_0 + \varepsilon_i$$

$$\varepsilon_i \sim N(0, \sigma^2)$$

Two sample t-test  
(equal variance)

$$Y_i \quad i=1, \dots, m+n$$

$$G_i = \begin{cases} 0 & \text{obs. } i \text{ from pop. 1} \\ 1 & \text{" " " 2} \end{cases}$$

t-test on the coefficient  $\beta_1$

$$Y_i = \beta_0 + \beta_1 G_i + \varepsilon_i$$

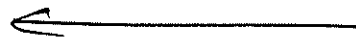
Paired t-test

$$Y_{ij} = \beta_0 + \alpha_j + \beta_1 G_{ij} + \varepsilon_{ij}$$

$$i=1, 2 \quad \alpha_j \sim N(\mu)$$

$$j=1, \dots, n \quad 16$$

t-test



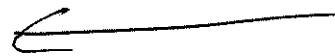
Regression  
ST552

Paired



Mixed Models  
ST553  
ST555

Binomial Proportions



Logistic Regression  
ST623

Multinomial



Generalized  
Linear Model

