## infer

## An R package for tidy statistical inference

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2018-01-27 (R User Day at Data Day Texas)

## Understanding who you are

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- Who uses hypothesis testing/confidence intervals at least once a week?
- Who uses the tidyverse at least once a week?
- Who has heard of permutation testing?
  - Randomization-based methods?
  - Resampling methods?
  - o Bootstrap methods?

## Pre-requisites for this talk

• Some experience with statistical inference (hypothesis testing / confidence intervals)

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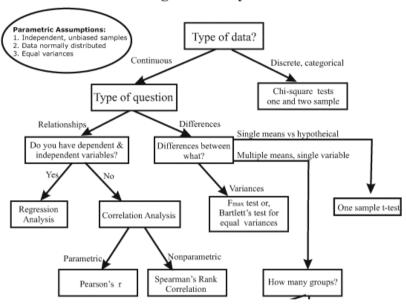
- Some experience with statistical inference (hypothesis testing / confidence intervals)
- A admiration, abundance of love, won't do anything without it respect for the tidyverse and its power to get more users into doing data analysis/visualization quickly
  - The pit of success

## Pre-requisites for this talk

- Some experience with statistical inference (hypothesis testing / confidence intervals)
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  - The pit of success
- Ability Desire to think differently about statistical inference using computational methods as the driver

#### Is this statistical inference to you?

#### Flow Chart for Selecting Commonly Used Statistical Tests



Students at Virginia Tech studied which vehicles come to a complete stop at an intersection with four-way stop signs, selecting at random the cars to observe. The explanatory variable used here is the arrival position of vehicles approaching an intersection all traveling in the same direction. They classified this arrival pattern into three groups: whether the vehicle arrives alone (single), is the lead in a group of vehicles, or is a follower in a group of vehicles. Is there an association between arrival pattern and whether a complete stop or not\_complete was made?

- From "Introduction to Statistical Investigations" by Tintle et al.

## Which type of hypothesis test should we conduct here?

- A. Independent samples t-test
- B. One proportion test
- C. Chi-Square test of independence
- D. ANOVA

```
# A tibble: 10 x 2
   stop_type
                vehicle_type
   <chr>>
                <chr>
 1 complete
                single
 2 complete
                follow
 3 complete
                lead
 4 complete
                single
 5 not_complete follow
 6 not_complete lead
 7 complete
                single
 8 complete
                follow
 9 complete
                single
10 complete
                lead
```

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#### Answer:

• C. Chi-Square Test of Independence

• Using a data argument

```
chisq.test(data = car_stop, x = stop_type, y = vehicle_type)
```

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```
chisq.test(data = car_stop, x = stop_type, y = vehicle_type)

Error in chisq.test(data = car_stop, x = stop_type,
    y = vehicle_type)
```

• Using a data argument

```
chisq.test(data = car_stop, x = stop_type, y = vehicle_type)

Error in chisq.test(data = car_stop, x = stop_type,
    y = vehicle_type)
```

• Using a formula

```
chisq.test(data = car_stop,
    formula = vehicle_type ~ stop_type)
```

• Using a data argument

```
chisq.test(data = car_stop, x = stop_type, y = vehicle_type)

Error in chisq.test(data = car_stop, x = stop_type,
    y = vehicle_type)
```

• Using a formula

## Finally

```
chisq.test(car_stop$stop_type, car_stop$vehicle_type)
```

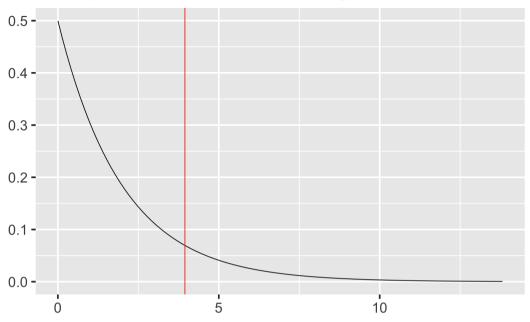
```
data: car_stop$stop_type and car_stop$vehicle_type
X-squared = 3.9476, df = 2, p-value = 0.1389
```

Pearson's Chi-squared test

#### ?chisq.test()



#### Chi-square distribution with 2 degrees of freedom



## P-value is 0.1389.

Is there an association between arrival pattern and whether or not a complete stop was made?

The null hypothesis

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# The null hypothesis

No association exists between the arrival vehicle's position and whether or not it makes a complete stop.

## The alternative hypothesis

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## The null hypothesis

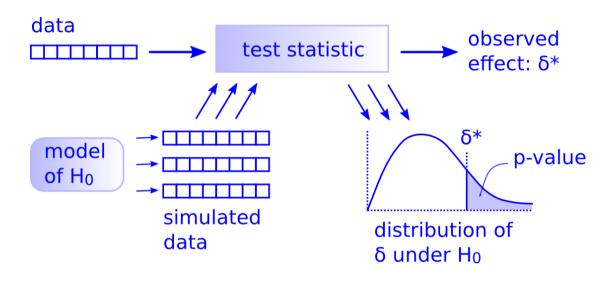
No association exists between the arrival vehicle's position and whether or not it makes a complete stop.

## The alternative hypothesis

An association exists between the arrival vehicle's position and whether or not it makes a complete stop.

How can computation help us to understand what is going on here?

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# The tricky step

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## Modeling the null hypothesis

• How do we simulate data assuming the null hypothesis is true in our problem (there is no association between the variables)?

## The tricky step

#### Modeling the null hypothesis

- How do we simulate data assuming the null hypothesis is true in our problem (there is no association between the variables)?
- What might the sample data look like if the null was true?

#### Properties of the original sample collected

#### car\_stop %>% count(stop\_type, vehicle\_type)

```
# A tibble: 6 x 3
           vehicle_type
 stop_type
                             n
        · <chr>
 <chr>
                          <int>
1 complete follow
                            76
2 complete lead
                            38
3 complete
             single
                            151
4 not_complete follow
                            22
5 not_complete lead
                             5
6 not_complete single
                            25
```

#### Properties of the original sample collected

```
car_stop %>% count(stop_type, vehicle_type)
# A tibble: 6 x 3
  stop_type
               vehicle_type
                                 n
           <chr>
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1 complete follow
                                76
2 complete lead
                                38
3 complete
               single
                               151
4 not_complete follow
                                22
5 not complete lead
                                  5
6 not_complete single
                                25
 ( orig_table <- car_stop %>%
     janitor::tabyl(stop type, vehicle type) )
    stop_type follow lead single
     complete
                   76
                        38
                              151
 not_complete
                   22
                         5
                               25
                  Slides available at http://bit.ly/infer-austin
                  Package webpage at https://infer.netlify.com
```

#### Permute the sample data

```
# A tibble: 317 x 2
                vehicle_type
   stop_type
   <fct>
                <fct>
 1 complete
                follow
 2 not_complete follow
 3 complete
                follow
 4 not_complete single
 5 complete
                single
 6 not_complete follow
 7 not complete single
 8 complete
                follow
 9 not_complete lead
10 complete
                lead
# ... with 307 more rows
```

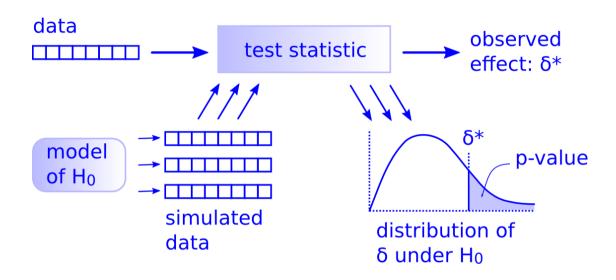
#### Permute the sample data

```
# A tibble: 317 x 2
                vehicle_type
   stop_type
   <fct>
                <fct>
                follow
1 complete
2 not_complete follow
3 complete
                follow
4 not_complete single
5 complete
                single
6 not_complete follow
7 not_complete single
8 complete
                follow
9 not_complete lead
10 complete
                lead
# ... with 307 more rows
    stop_type follow lead single
     complete
                       35
                  79
                              151
not_complete
                  19
                               25
                        8
```

## Comparing the original and permuted sample

```
orig table %>% janitor::adorn totals(where = c("row", "col"))
   stop_type follow lead single Total
    complete
                 76
                     38
                            151
                                  265
not_complete
                 22
                    5
                            25
                                  52
       Total
                 98
                     43
                            176
                                  317
new table %>% janitor::adorn totals(where = c("row", "col"))
   stop_type follow lead single Total
    complete
                79
                     35
                            151
                                  265
not_complete
                 19
                    8
                            25
                                  52
       Total
                 98
                    43
                           176
                                 317
```

#### Where are we?



#### Test statistic

- Chi-square test statistic (Wikipedia)
  - Measure of how far what we observed in our sample is from what we would expect if the null hypothesis was true

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- Chi-square test statistic (Wikipedia)
  - Measure of how far what we observed in our sample is from what we would expect if the null hypothesis was true

chisq.test(car\_stop\$stop\_type, car\_stop\$vehicle\_type)\$statistic

X-squared 3.947648

#### For the permuted data

chisq.test(perm1\$stop\_type, perm1\$vehicle\_type)\$statistic

X-squared 1.408986

### For the permuted data

chisq.test(perm1\$stop\_type, perm1\$vehicle\_type)\$statistic

X-squared 1.408986

#### Another permutation

chisq.test(perm2\$stop\_type, perm2\$vehicle\_type)\$statistic

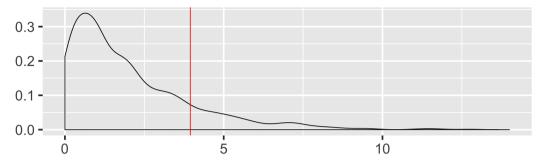
X-squared 0.3604528

## What does the distribution of multiple repetitions of the permuted data look like?

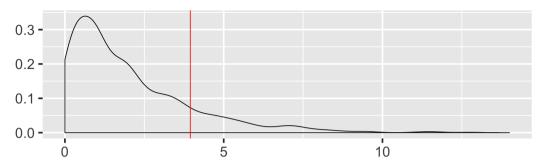
```
# A tibble: 1,000 x 2
   replicate stat
   <fct>
             <dbl>
             1.05
1 1
2 2
          7.24
            0.253
            2.16
            1.60
5 5
            3.95
            1.94
8 8
            1.68
            0.242
9 9
10 10
             3.26
# ... with 990 more rows
```

The distribution of multiple repetitions of the permuted data

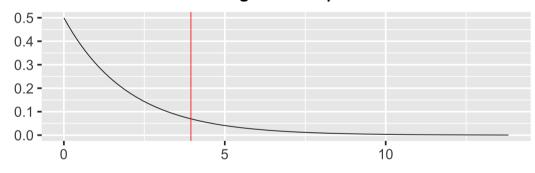
#### The distribution of multiple repetitions of the permuted data



#### The distribution of multiple repetitions of the permuted data



#### Recall the traditional method using the Chi-square distribution



Slides available at http://bit.ly/infer-austin Package webpage at https://infer.netlify.com

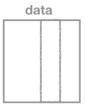
#### Objectives of infer

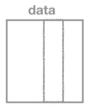
• Implement common classical inferential techniques in a tidyverse-friendly framework that is expressive of the underlying procedure.

#### Objectives of infer

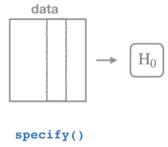
- Implement common classical inferential techniques in a tidyverse-friendly framework that is expressive of the underlying procedure.
  - Dataframe in, dataframe out
  - Compose tests and intervals with pipes
  - Unite computational and approximation methods
  - Reading a chain of infer code should describe the inferential procedure

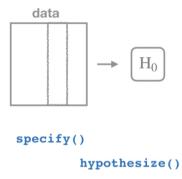
# The infer verbs

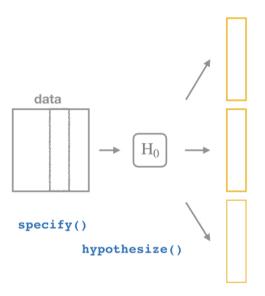


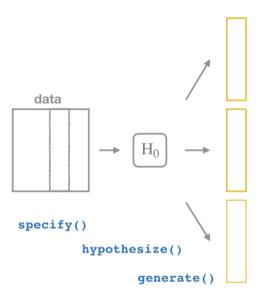


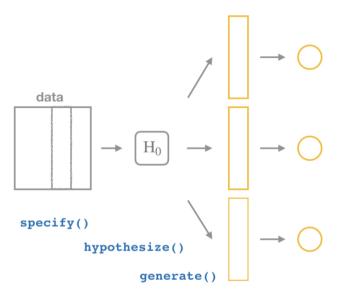
specify()

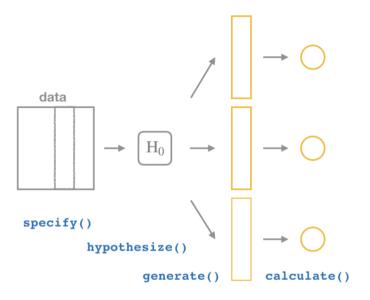


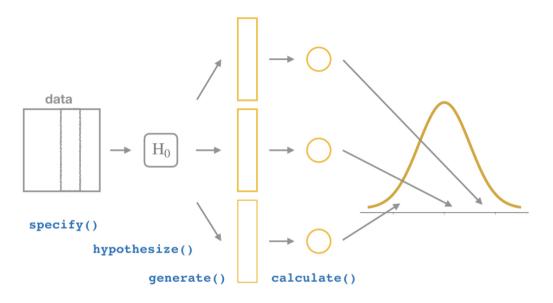


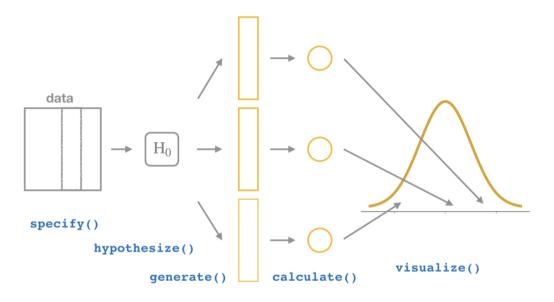




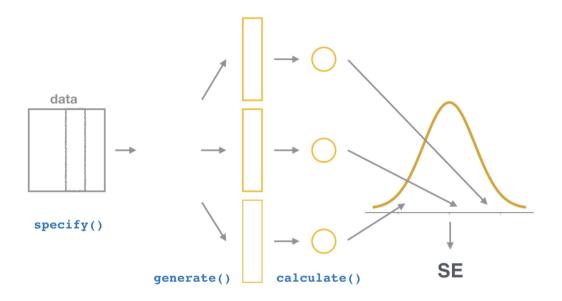








#### Confidence Interval

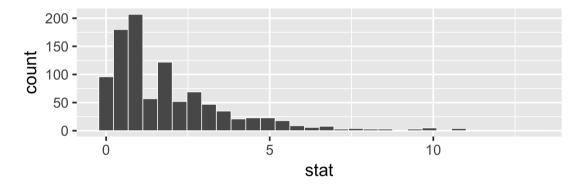


```
car_stop %>%
  specify(stop_type ~ vehicle_type) %>%
  hypothesize(null = "independence") %>%
  generate(reps = 1000, type = "permute") %>%
  calculate(stat = "Chisq")
```

```
# A tibble: 1,000 x 2
   replicate stat
   <fct>
             <dbl>
 1 1
             0.509
 2 2
             0.760
 3 3
             1.79
             1.83
 5 5
             0.525
 6 6
             4.74
             1.11
 7 7
 8 8
             2.51
 9 9
             1.11
10 10
             0.421
# ... with 990 more rows
```

#### Back to the example

```
car_stop %>%
  specify(stop_type ~ vehicle_type) %>%
  hypothesize(null = "independence") %>%
  generate(reps = 1000, type = "permute") %>%
  calculate(stat = "Chisq") %>%
  visualize()
```



#### What's to come

- Wrapper functions: t\_test, chisq\_test, etc.
- Generalized input to calculate()
  - For example, calculate(trimmed\_mean)
  - Support for more advanced regression models
- Adding features to visualize()
  - Show both traditional and computation methods
- Implement list-columns in the generate() step

#### Tips and tricks for package development

- Use GitHub and pull requests to the master branch
- Create useful vignettes so others know how your pkg works
- Write tests and assertions for your code
  - Buy and read Richie's Testing R Code book
- Let travis-ci do the work for you
- Use Hadley's pkgdown package to build a pkg website
  - Host it on Netlify.com to be super cool

#### More info

- https://infer.netlify.com
  - Many examples under Articles there with more to come
  - Plans to be implemented in www.ModernDive.com by this summer
    - Sign up for the ModernDive mailing list for details
- Two DataCamp courses currently launched that use infer
  - Inference for Numerical Data by Mine Cetinkaya-Rundel
  - Inference for Regression by Jo Hardin
- Two more DataCamp courses to be launched

- Special thanks to Andrew Bray and the other pkg contributors
- Slides created via the R package xaringan by Yihui Xie
- Slides available at http://bit.ly/infer-austin
- Source code for these slides at https://github.com/ismayc/talks/tree/master/data-daytexas-infer

