

# Lab 1 Part 1 - Foundational Exercises

Team03: Savita Chari, Tymon Silva, Denny Lehman

## Professional Magic

Your aunt (who is a professional magician), claims to have created a pair of magical coins that share a connection to each other that makes them land in the same way. The coins are always flipped at the same time. For a given flip  $i \in \{1, 2, 3, \dots\}$ , let  $X_i$  be a Bernoulli random variable representing the outcome of the first coin, and let  $Y_i$  be a Bernoulli random variable representing the outcome of the second coin. You assume that each flip of the pair is independent of all other flips of the pair. You also assume that

$$P(X_i = 0) = P(X_i = 1) = P(Y_i = 0) = P(Y_i = 1) = 1/2,$$

and write,

$$P(X_i = Y_i) = p.$$

Your aunt claims that  $p > 1/2$ .

You design a test to evaluate your aunt's claim. You flip the coins 3 times and define your test statistic to be the sum  $X_1 + Y_1 + X_2 + Y_2 + X_3 + Y_3$

Your null hypothesis is that  $p = 1/2$ . You plan to reject the null if your test statistic is 0 or 6.

1. What is the type 1 error rate of your test?
2. What is the power of your test for the alternate hypothesis that  $p = 3/4$ ?

## 1. What is the type 1 error rate of the test?

The type 1 error occurs when we reject the null hypothesis when it is actually true. Type I error, a false positive, is typically denoted as alpha

$$\alpha = P(\text{reject } H_0 | H_0)$$

We are given that  $H_0$  is when  $p = 1/2$ .  $p$  is the probability that  $X_i = Y_i$

$$H_0 : p = \frac{1}{2}, p = P(X_i = Y_i)$$

Let's define a trial as flipping both coins at the same time, where coin X has outcomes 0 and 1 and coin Y has 0 and 1

$$X \in \{0, 1\}$$

$$Y \in \{0, 1\}$$

$\theta_1$  = outcome of the event, which is two coin flips

Event space of  $\theta_1$

$$\theta_1 \in \{\{0, 0\}, \{0, 1\}, \{1, 0\}, \{1, 1\}\}$$

$$P(X = 0) = \frac{1}{2}$$

$$P(Y = 0) = \frac{1}{2}$$

$$P(X = 1) = \frac{1}{2}$$

$$P(Y = 1) = \frac{1}{2}$$

$$P(X = x, Y = y) = P(X = x) * P(Y = y)$$

We know that each flip of the pair has the following potential outcomes:

$$P(X_i = Y_i) = P(X_i = 0, Y_i = 0) + P(X_i = 1, Y_i = 1)$$

$$P(X_i = Y_i | H_o) = 0.5 = P(X_i = 0, Y_i = 0) + P(X_i = 1, Y_i = 1)$$

$$P(X_i = 0, Y_i = 0) = \frac{1}{4}$$

$$P(X_i = 1, Y_i = 1) = \frac{1}{4}$$

We are also given that our test statistic,  $\theta_{\text{hat}}$ , will be equal to

$$\hat{\theta}_n = \sum_{i=1}^n X_i + Y_i$$

$$\hat{\theta}_3 = X_1 + Y_1 + X_2 + Y_2 + X_3 + Y_3$$

From the problem definition, we reject the null when our test statistic  $\theta$  is 0 or 6 after 3 flips of pairs of coins. Type I error is thus

$$\alpha = P(\text{reject } H_o | H_o)$$

$$\alpha = P(\hat{\theta}_3 \in \{0, 6\} | p = \frac{1}{2})$$

$$\alpha = P(\hat{\theta}_3 = 0 | p = \frac{1}{2}) + P(\hat{\theta}_3 = 6 | p = \frac{1}{2})$$

Solve for each term using the binomial distribution formula

binomial distribution formula

$$P_x = nCx * q^x (1 - q)^{n-x}$$

$q$  was chosen to avoid confusion with  $p$  from the problem statement. Next, solve for  $P(\hat{\theta}_3 = 0 | p = \frac{1}{2})$  where  $q$  is  $P(X_i = 0, Y_i = 0) = 1/4$

$\hat{\theta}_3 = 0$  means that after 3 trials, the test statistic has value of 0 (all 0's)

$$P(\hat{\theta}_3 = 0 | p = \frac{1}{2}) = 0C3q^3 * (1 - q)^0$$

$$P(\hat{\theta}_3 = 0 | p = \frac{1}{2}) = (1)(1/4)^3(1)$$

$$P(\hat{\theta}_3 = 0 | p = \frac{1}{2}) = (1/4)^3$$

solve for  $P(\hat{\theta}_3 = 6|p = \frac{1}{2})$

where  $q$  is  $P(X_i = 1, Y_i = 1) = 1/4$

$\hat{\theta}_3 = 6$  means after all three trials, the test statistic has the value of 6 (all 1's)

$$P(\hat{\theta}_3 = 6|p = \frac{1}{2}) = 3C3q^3 * (1 - q)^0$$

$$P(\hat{\theta}_3 = 6|p = \frac{1}{2}) = (1)(1/4)^3(1)$$

$$P(\hat{\theta}_3 = 6|p = \frac{1}{2}) = (1/4)^3$$

now substitute back in for alpha

$$\alpha = P(\hat{\theta}_3 = 0|H_o) + P(\hat{\theta}_3 = 6|H_o)$$

$$\alpha = (\frac{1}{4})^3 + (\frac{1}{4})^3$$

$$\alpha = \frac{2}{64} = \frac{1}{32}$$

## 2. What is the power of your test for the alternate hypothesis that $p = 3/4$ ?

The power of our test is the Probability that we have successfully rejected the null hypothesis when the alternative is true.  $Power = 1 - \beta$  Where  $\beta$  is the type 2 error, or the false negative rate. Type 2 error occurs when we fail to reject the null hypothesis when we should.

$$H_a : p = \frac{3}{4}, p = P(X_i = Y_i)$$

$$Power = P(rejectnull|H_a)$$

We assume that the likelihood for heads and tails remains the equal, but the likelihood of getting  $X_i = Y_i$  changes to  $\frac{3}{4}$ . Therefore,

$$P(X_i = Y_i|H_a) = 3/4$$

$$P(X_i = Y_i) = P(X_i = 0, Y_i = 0) + P(X_i = 1, Y_i = 1)$$

$$P(X_i = 0, Y_i = 0) = P(X_i = 1, Y_i = 1)$$

$\therefore$

$$P(X_i = 0, Y_i = 0) = 3/8$$

$$P(X_i = 1, Y_i = 1) = 3/8$$

$$P(X_i = 0, Y_i = 1) = 1/8$$

$$P(X_i = 1, Y_i = 0) = 1/8$$

When fail to reject

$$\hat{\theta}_3 = X_1 + Y_1 + X_2 + Y_2 + X_3 + Y_3$$

reject  $H_o$  when  $\hat{\theta}_3 = 0$  or  $\hat{\theta}_3 = 6$

$$Power = P(rejectnull|H_a)$$

$$Power = P(\hat{\theta}_3 \in \{0, 6\}|p = \frac{3}{4})$$

$$Power = P(\hat{\theta}_3 = 0|p = 3/4, q = 3/8) + P(\hat{\theta}_3 = 6|p = 3/4)$$

Now we solve for the two terms

$$\begin{aligned} P(\hat{\theta}_3 = 0|p = 3/4) &= 3C0 * (3/8)^3 * (1 - 3/8)^0 \\ &= (1)(3/8)^3(1) \\ &= \frac{27}{512} \\ P(\hat{\theta}_3 = 6|p = 3/4) &= 3C3 * (3/8)^3 * (1 - 3/8)^0 \\ &= (1)(3/8)^3(1) \\ &= \frac{27}{512} \end{aligned}$$

$$\begin{aligned} Power &= P(\hat{\theta}_3 = 0|p = 3/4) + P(\hat{\theta}_3 = 6|p = 3/4) \\ Power &= \left(\frac{3}{8}\right)^3 + \left(\frac{3}{8}\right)^3 \\ Power &= 2 * \left(\frac{3}{8}\right)^3 \\ Power &\approx 0.105 \end{aligned}$$

## Wrong Test, Right Data

*Our Analysis :*

The response categories in Likert scales creates rank order/ordinal data, but the intervals between their values cannot be compared or presumed equal. Therefore, standard statistical analysis such as the mean (central tendency), standard deviation, etc. are meaningless and inappropriate with ordinal data. Hence, this data is not suitable for a paired t-test. The output of this test would be the difference in means from the likability of the mobile website vs the regular website. While it is possible to draw conclusions in the directionality of the result of a paired t-test on this data, we cannot interpret the magnitude of the outcome because we are using ordinal data and taking the difference of their means is meaningless. In conclusion, the paired t-test could answer which version of the website is more likable, but not by how much.

## Test Assumptions

For the four following questions, your task is to evaluate the assumptions for the given test using your background knowledge and examining the data.

### World Happiness

You would like to know whether people in countries with high GDP per capita (higher than the mean) are more happy or less happy than people in countries with low GDP (lower than the mean). List all assumptions for a two-sample t-test and evaluate them.

## Two-sample t-test assumptions

1. Metric variables

The Life Ladder variable is a metric variable. We can justify this by reviewing the FAQ for this dataset that states, "... it asks respondents to think of a ladder, with the best possible life for them being a 10, and the worst possible life being a 0. They are then asked to rate their own current lives on that 0 to 10 scale." From this description, we know the responses fall on a 0 to 10 scale, where the distances or gaps between each point on the scale are the same, just like the rungs on a ladder are equidistant to each other.

2. IID

Here, we assume IID is upheld since Gallup <sup>1</sup> is a reputable research firm, and we expect their sampling and data collection processes maintain the independence and identically distributed assumption.

3. Data is normally distributed (check sample size)

Since we have sufficient sample sizes (greater than 30) in each of the Low (n=105) and High (n=121) groups for GDP, we can assume normality by the Central Limit Theorem.

Thus, all the assumptions for a two-sample t-test are valid and the test would be appropriate.

## Study question 3: Legislator age

You would like to test whether Democratic or Republican senators are older. List all assumptions for a Wilcoxon rank-sum test and evaluate them.

Using Wilcoxon rank-sum test we have been asked to run a statistical test using the data provided and maintained by GovTrack, ProPublica, MapLight, FiveThirtyEight etc. The correctness of id's which relate the record to other databases, name information (first, last, etc.), biographical information (birthday, gender), and terms served in Congress guaranteed. <sup>2</sup>

## Test Assumption:

The database maintains information about all the members of the United States Congress, Senators, Presidents and Vice Presidents of the United States. The study question focuses our test on Democratic and Republican *senators*, therefore we will limit the study only to the senators of USA. The field *type = 'sen'* filters only desired rows of data from this dataset.

## Hypothesis

**H<sub>0</sub>: Probability that the Democrat Senators are older than Republicans is same as the probability of Republican Senators being older than their Democrat counterpart.**

## What is Wilcoxon rank-sum test?

Wilcoxon rank-sum test is a non-parametric, distribution-free test for two independent samples. Though it involves assumptions, but those assumptions are less restrictive than the assumptions for parametric tests. It considers the ranks instead of the metric value of the variable. It uses the order of variables to construct statistics that can be used to test hypothesis.

*Advantage* : The population distribution doesn't have to be normal, so it's easier to justify a rank-based test. It is a good choice of test for smaller sample size.

*Disadvantage* : Since these tests do not use the metric information they lose statistical power.

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<sup>1</sup><https://www.gallup.com/corporate/212381/who-we-are.aspx>

<sup>2</sup><https://github.com/unitedstates/congress-legislators>

**The Wilcox rank-sum test works on following principals:**

1. The samples are interval scale and list the score from lowest to highest. Higher rank gets higher score
2. Only considers rank instead of looking at metric value of the variable
3. Uses order of variables to construct statistics that is used for hypothesis

**The Wilcox rank-sum test makes the following assumptions:**

1. Test involves independent, unpaired samples
2. The data is IID
3. The Data is Ordinal Variable

## Data and Methodology

**In order to perform Wilcoxon test we need 2 types of columns from our dataset**

1. [X] A categorical column with 2 distinct groups : We chose the ‘party’ column for this requirement but the data set includes more than two categories. We filtered only the Republican and Democrat data in order to perform Wilcoxon rank-sum test as we are interested in only those categories. Thus we fulfill the requirement of having only 2 categories.
2. [Y] A statistical column with numeric outcome: We chose the ‘birthday’ field for this requirement. Though this field is of class ‘Date’, with some manipulation we can make it a metric outcome

```
## [1] "The Dataset had 98 Rows and 4 Columns"
```

### Data Transformation

As mentioned above, the birthday field is of ‘Date’ class and is in YYYY/MM/DD format. In order to find out the age of the Senator we calculate the current age of the Senator by doing a datediff between current date and the birthday

```
## # A tibble: 3 x 5
##   full_name      birthday party    type AgeInYears
##   <chr>          <date>    <chr>   <chr>      <dbl>
## 1 Sherrod Brown  1952-11-09 Democrat sen         69
## 2 Maria Cantwell 1958-10-13 Democrat sen         63
## 3 Benjamin L. Cardin 1943-10-05 Democrat sen         78
```

### Exploring Assumption 1 : Are the data sets independent?

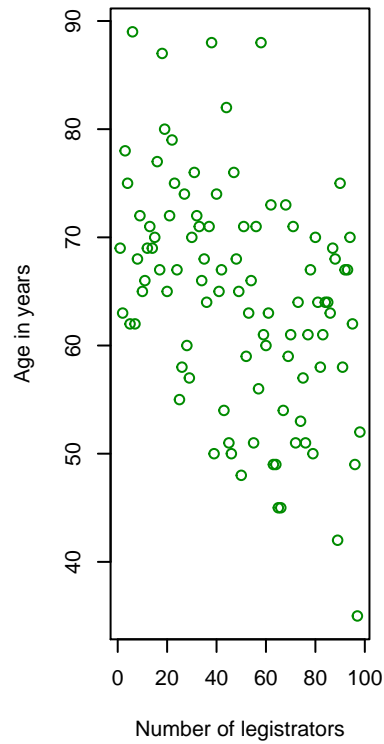
Well, yes, a Senator cannot belong to 2 parties simultaneously, so the data is independent. As mentioned in the introduction section, the source of the data is the congress-legislators project. Data such as information (first, last, etc.), biographical information (birthday, gender) of every members of the United States Congress (1789-Present) is maintained. Data of one Senator does not depend on the other. So we can safely assume the data is independent. In this test we are using the entire population of data available to us.

### Exploring Assumption 2 : Is the population IID/Ordinal?

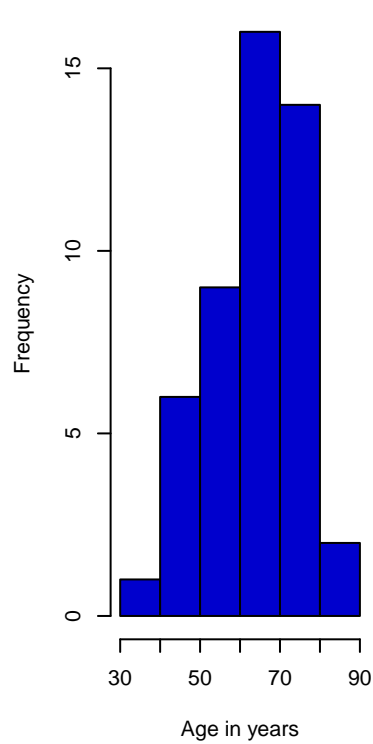
From the graph below we can tell that the age distribution between republicans and democrats is somewhat normal, and not too skewed. The data is IID. Moreover, as per Central Limit Theorem (CMT), when we have a sufficiently large sample size (98 in our case), the sampling distribution starts to approximate a normal distribution.

*Given this information, Wilcoxon rank – sum test is not a good choice for this use case*

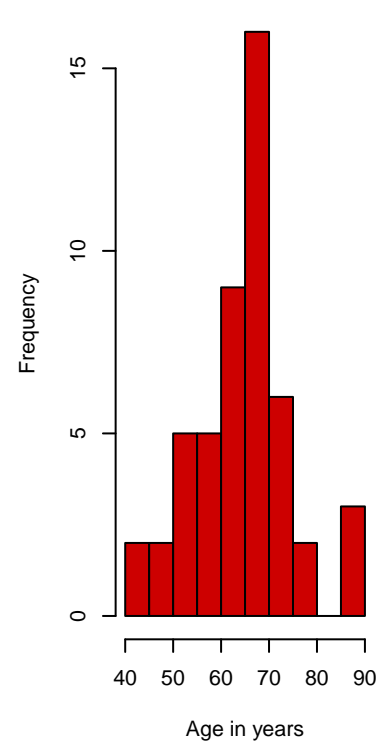
**Age distribution of all the Senat**



**Age distribution of Democrats**

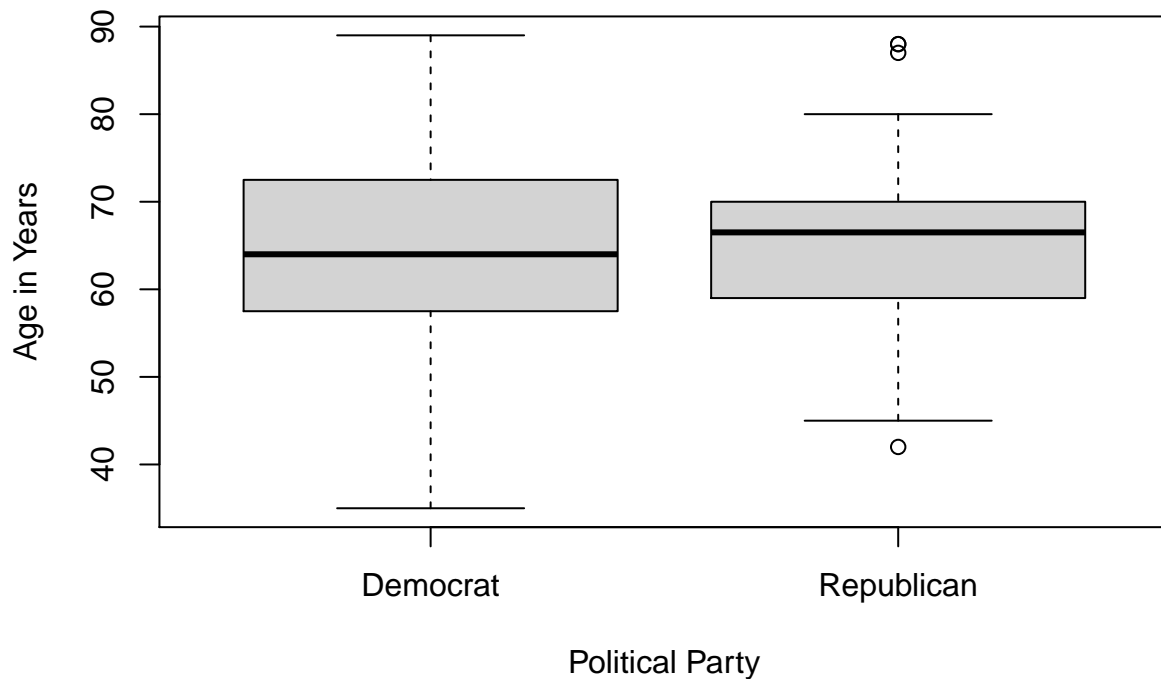


**Age distribution of Republican**



Also, from the box plot below, it is very clear that we can calculate variance on the metric data (AgeInYears). Democrats have a larger variance than the Republicans. In the Republican data set there are some outliers too.

## Variance in the Age



### Performing two sided Wilcoxon rank-sum test

In order to prove the null hypothesis we have been asked to perform the Wilcoxon rank-sum test. Our data satisfies the requirements for conducting this test.

```
wilcox.test(legislator_DemRep_Age_data$AgeInYears~legislator_DemRep_Age_data$party, mu=0,
            alt = "two.sided", conf.int = T, conf.level=0.95, paired=F, exact=F, correct=T)
```

```
##
## Wilcoxon rank sum test with continuity correction
##
## data:  legislator_DemRep_Age_data$AgeInYears by legislator_DemRep_Age_data$party
## W = 1193.5, p-value = 0.966
## alternative hypothesis: true location shift is not equal to 0
## 95 percent confidence interval:
##  -4.000097  4.000001
## sample estimates:
## difference in location
##      -7.858072e-05
```

With p-value being greater than 0.05, We cannot reject the null hypothesis. Which stated that the probability that the Democrat Senators are older than Republicans is same as the probability of Republican Senators being older than their Democrat counterpart.

### Observation

It is our opinion that in this scenario where we have an option to run either parametric or non-parametric test, we should choose parametric t-test because it has more statistical power as compared to non-parametric test such as wilcoxon Rank-Sum test.

Our data fulfills all the requirements to run a t-test



1. The samples are independent of one another and are Metric scale.
2. The populations have equal variance or spread
3. The populations are normally distributed (IID)

```
t.test(legislator_DemRep_Age_data$AgeInYears~legislator_DemRep_Age_data$party)

##
##  Welch Two Sample t-test
##
## data:  legislator_DemRep_Age_data$AgeInYears by legislator_DemRep_Age_data$party
## t = -0.30936, df = 95.063, p-value = 0.7577
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  -4.858245  3.548245
## sample estimates:
##  mean in group Democrat mean in group Republican
##                64.125                64.780
```

**With p-value being greater than 0.05, We cannot reject the null hypothesis.** Which stated that the probability that the Democrat Senators are older than Republicans is same as the probability of Republican Senators being older than their Democrat counterpart.

The t-test further observes that the mean age of Republican Senators is approximately 6 months higher than their Democratic counterpart.

### Test Outcome and it's correctness

1. We used the mean to create our test statistics which is well accepted distributions for statistical testing.
2. We followed the decision rules of a hypothesis test which gives us the guarantee that our false positive rejection rate (the type 1 error rate) is bounded.
3. Our test sample size is >30 so it meets the standard t-test heuristic.

### Conclusion for Study question 3: Legislator age

It is important to observe the data before choosing the right kind of test to make correct predictions and to leverage the full power of statistics.

### It's for your health!

You would like to use it [wine dataset] to test whether countries have more deaths from heart disease or from liver disease. List all assumptions for a signed-rank test and evaluate them.

```
##      country alcohol deaths heart liver
## 1  Australia      2.5    785   211  15.3
## 2   Austria      3.9    863   167  45.6
## 3  Belg/Lux      2.9    883   131  20.7
## 4   Canada      2.4    793   191  16.4
## 5   Denmark      2.9    971   220  23.9
## 6   Finland      0.8    970   297  19.0
## 7    France      9.1    751    71  37.9
## 8   Iceland      0.8    743   211  11.2
## 9    Ireland      0.7   1000   300   6.5
## 10   Israel      0.6    834   183  13.7
## 11    Italy      7.9    775   107  42.2
```

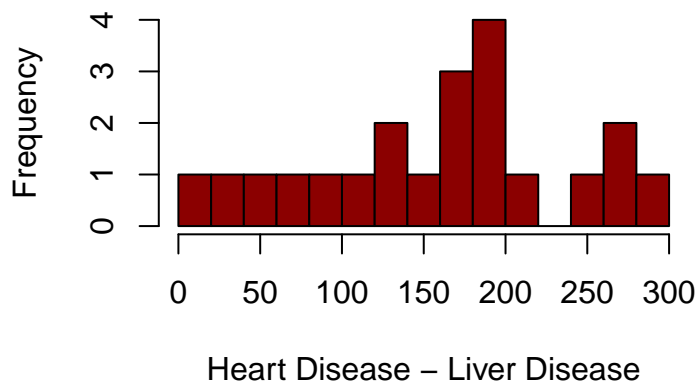
## 12	Japan	1.5	680	36	23.2
## 13	Netherlands	1.8	773	167	9.2
## 14	New Zealand	1.9	916	266	7.7
## 15	Norway	0.8	806	227	12.2
## 16	Spain	6.5	724	86	36.4
## 17	Sweden	1.6	743	207	11.2
## 18	Switzerland	5.8	693	115	20.3
## 19	UK	1.3	941	285	10.3
## 20	US	1.2	926	199	22.1
## 21	West Germany	2.7	861	172	36.7

### Wilcoxon signed-rank test assumptions

1. Metric variables We need metric variables for a paired Wilcoxon signed-rank test, and the heart and liver variables satisfy this assumption. The Heart and Liver variables represent the number of deaths from heart and liver disease in each country.
2. IID  
There are some concerns with IID for this data. Since there dataset only contains 21 countries, we do not know how these countries were drawn, particularly if they were randomly select. Furthermore, we do not know how the data was collected, which means the data draw could not have been identically distributed. At a glance, we can see that there aren't any African countries in the dataset even though Africa has the most countries of any continent. This is an indication that the data was not randomly sampled.
3. Difference is symmetric The last assumption requires the difference between pairs follows a symmetric distribution. If you look at the histogram below of the difference between heart and liver disease deaths, one could argue that the distribution is skewed left, which violates our difference is symmetric assumption.

Due to major concerns with assumptions 2 and 3, the Wilcoxon signed-rank test would not be a valid test to perform to answer the question.

### Histogram of Differences



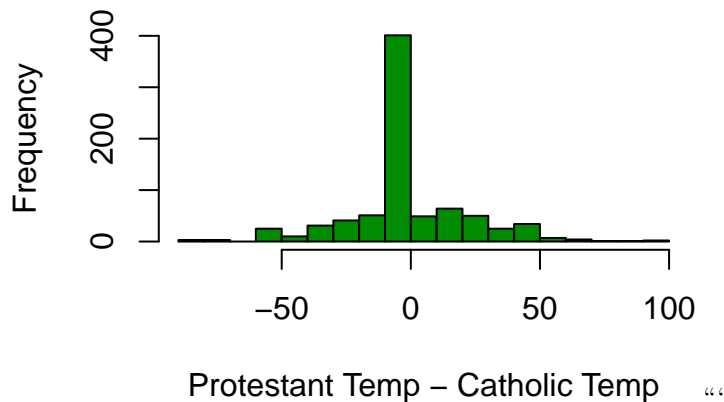
### Positive vibes

You would like to test whether the US population feels more positive towards Protestants or towards Catholics. List all assumption for a paired t-test and evaluate them.

```
## Warning: Missing column names filled in: 'X1' [1]
```

```
## # A tibble: 802 x 5
##       X1  year   id prottemp cathtemp
##   <dbl> <dbl> <dbl>   <dbl>   <dbl>
## 1     1     2004     4       50       50
## 2     2     2004     7        5       85
## 3     3     2004     9       50       50
## 4     4     2004    14       60       60
## 5     5     2004    21      100        0
## 6     6     2004    24        5        5
## 7     7     2004    25       50       60
## 8     8     2004    28       40       60
## 9     9     2004    29       50       50
## 10    10     2004    31       50       50
## # ... with 792 more rows
```

## Histogram of Differences



### Paired two-sample t-test assumptions

1. Metric variables

The prottemp and cathtemp variables are metric since they fall on 0 to 100 scales. In fact, respondents are told to think of the scale as a thermostat. And just like the degrees on a thermostat are equidistant for one another, we can argue the points on this 100-point scale are too. Therefore, we have metric variables.

2. IID

Again, we assume IID is upheld since the GSS<sup>3</sup> is a reputable survey, and we expect their sampling and data collection processes maintain the independence and identically distributed assumption.

3. Data is normally distributed

We do have a sample size of 802. So, by the Central Limit Theorem, we can assume the differences in our outcome variables will follow a normal distribution. We can see in the histogram below, that the distribution is indeed a tight, normal-looking distribution.

Thus, all the assumptions for a paired t-test are valid and the test would be appropriate to perform.

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<sup>3</sup><https://gss.norc.umd.edu/>