

An Investigation of Factors Influencing Emergency Healthcare Expenditures

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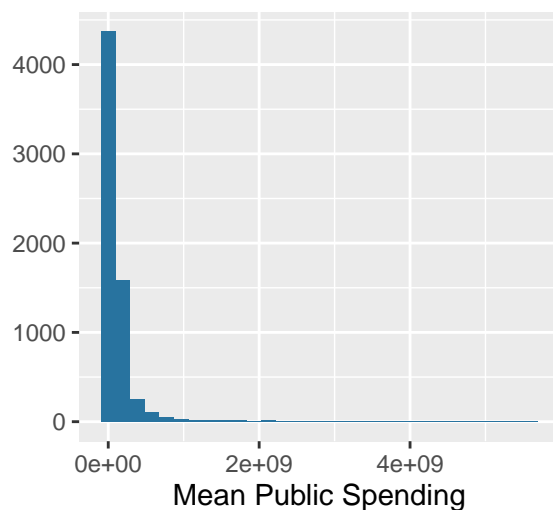
Abstract

Nature of the Data

The data includes summary of both male and female spending as “both”, so in order to perform an analysis on this data, we decided to exclude the both data points to avoid double counting? It is important to acknowledge that this data only included those who identified as either male or female, so this is not a complete representation of the population.

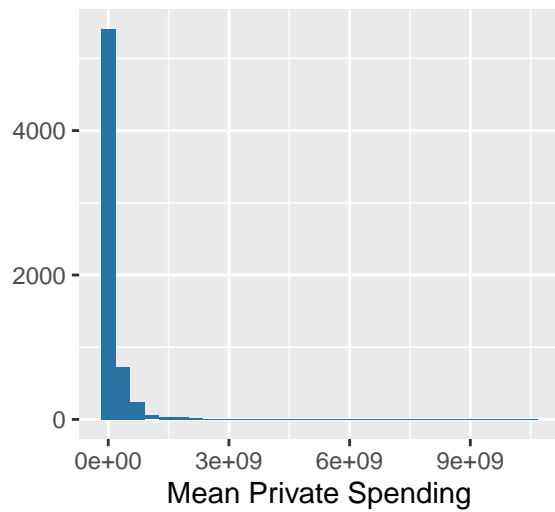
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

Normal Distribution of Mean Pub



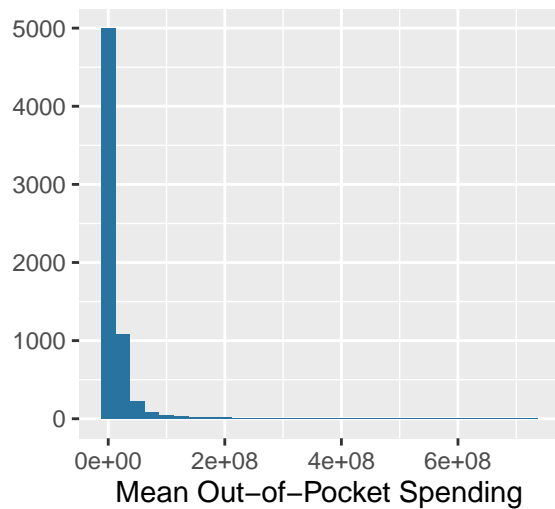
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

Normal Distribution of Mean Priv



```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

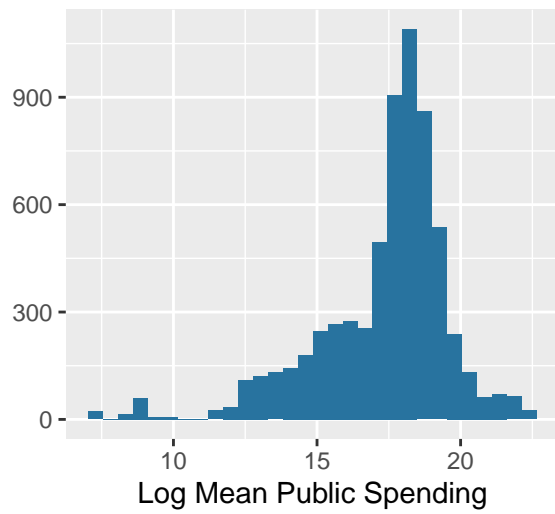
Normal Distribution of Mean Out



The normal distribution for public spending, private spending, and out-of-pocket pending all show a severe right skew in the data. Therefore, all three variables do not meet the normal distribution assumption needed for many tests, such as ANOVA; however, this can easily be resolved by applying a log transformation to the data to give a fairly normal distribution of the data.

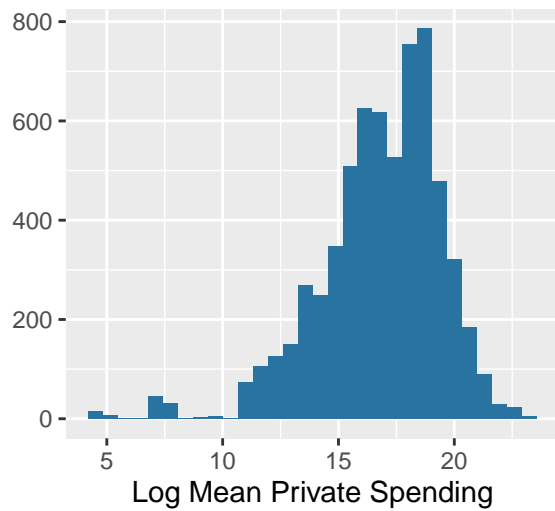
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

Log Normal Distribution of Mean



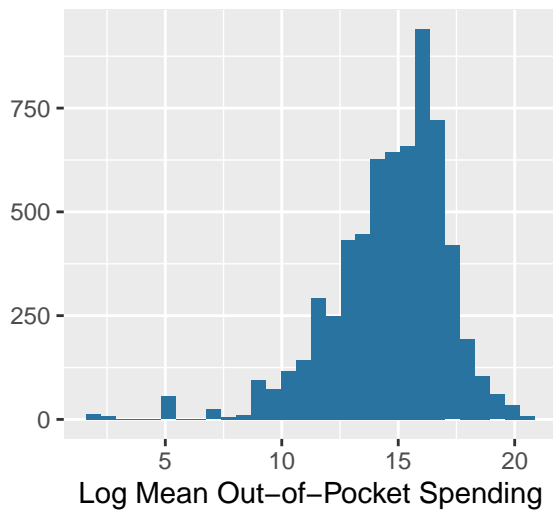
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

Log Normal Distribution of Mean



```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

Log Normal Distribution of Mean



These graphs of the log distribution of the various spending means appear to be fairly normal in distribution, which means they meet the requirements to be used in various analyses. In order to convert to log scale, those with mean_all, mean_pub, mean_pri, and mean_oop equal to zero must be excluded.

Gender

Our first question in this analysis is if males and females spend a different amount of money on emergency services.

First this t-test looks at overall differences in log mean emergency department spending between males and females

```
##
##  Welch Two Sample t-test
##
## data:  spending_malefemale$lmean_all by spending_malefemale$sex
## t = 1.4247, df = 6219.5, p-value = 0.1543
## alternative hypothesis: true difference in means between group Female and group Male is not equal to
## 95 percent confidence interval:
## -0.0315862  0.1996079
## sample estimates:
## mean in group Female    mean in group Male
##           18.06275           17.97874
```

This t-test shows that for mean spending of all emergency services payment types, the p value of 0.1543 (95% CI -0.0315862, 0.1996079) indicates there is not a significant difference between male and female spending.

Next, we perform a t-test on each type of insurance to see if there is a difference in spending between males and females:

```
##
##  Welch Two Sample t-test
##
## data:  spending_malefemale$lmean_pub by spending_malefemale$sex
## t = 1.8142, df = 6201, p-value = 0.0697
## alternative hypothesis: true difference in means between group Female and group Male is not equal to
## 95 percent confidence interval:
## -0.00833746  0.21532602
```

```
## sample estimates:
## mean in group Female    mean in group Male
##           17.40512           17.30162
```

The t-test on emergency services spending for people who have public insurance indicates there is not a significant difference between male and female spending, with p value of 0.0697 (95% CI -0.00833746, 0.21532602).

```
##
## Welch Two Sample t-test
##
## data:  spending_malefemale$lmean_pri by spending_malefemale$sex
## t = 0.70583, df = 6254.9, p-value = 0.4803
## alternative hypothesis: true difference in means between group Female and group Male is not equal to
## 95 percent confidence interval:
## -0.08283085  0.17603825
## sample estimates:
## mean in group Female    mean in group Male
##           16.82891           16.78231
```

The t-test on emergency services spending for people who have private insurance indicates there is not a significant difference between male and female spending, with p value of 0.4803 (95% CI -0.08283085, 0.17603825).

```
##
## Welch Two Sample t-test
##
## data:  spending_malefemale$lmean_oop by spending_malefemale$sex
## t = 0.9799, df = 6230.6, p-value = 0.3272
## alternative hypothesis: true difference in means between group Female and group Male is not equal to
## 95 percent confidence interval:
## -0.0615859  0.1846904
## sample estimates:
## mean in group Female    mean in group Male
##           14.66032           14.59877
```

The t-test on emergency services spending for people who pay out of pocket indicates there is not a significant difference between male and female spending, with p value of 0.3272 (95% CI -0.0615859, 0.1846904).

The t-tests for each type of insurance indicate that there is not enough evidence to reject the null hypothesis that emergency department spending is the same for males and females who have public insurance, private insurance, or pay out of pocket, leading us to the conclusion that gender does not influence emergency spending in the forms of payment studied here.

Disease category and Emergency Spending

In order to determine emergency department spending based on disease type, an ANOVA test is performed due to the data for spending on the log scale being normally distributed, relatively similar variance, and independent.

The null hypothesis for this ANOVA test is that the overall mean of spending are the same for each disease category

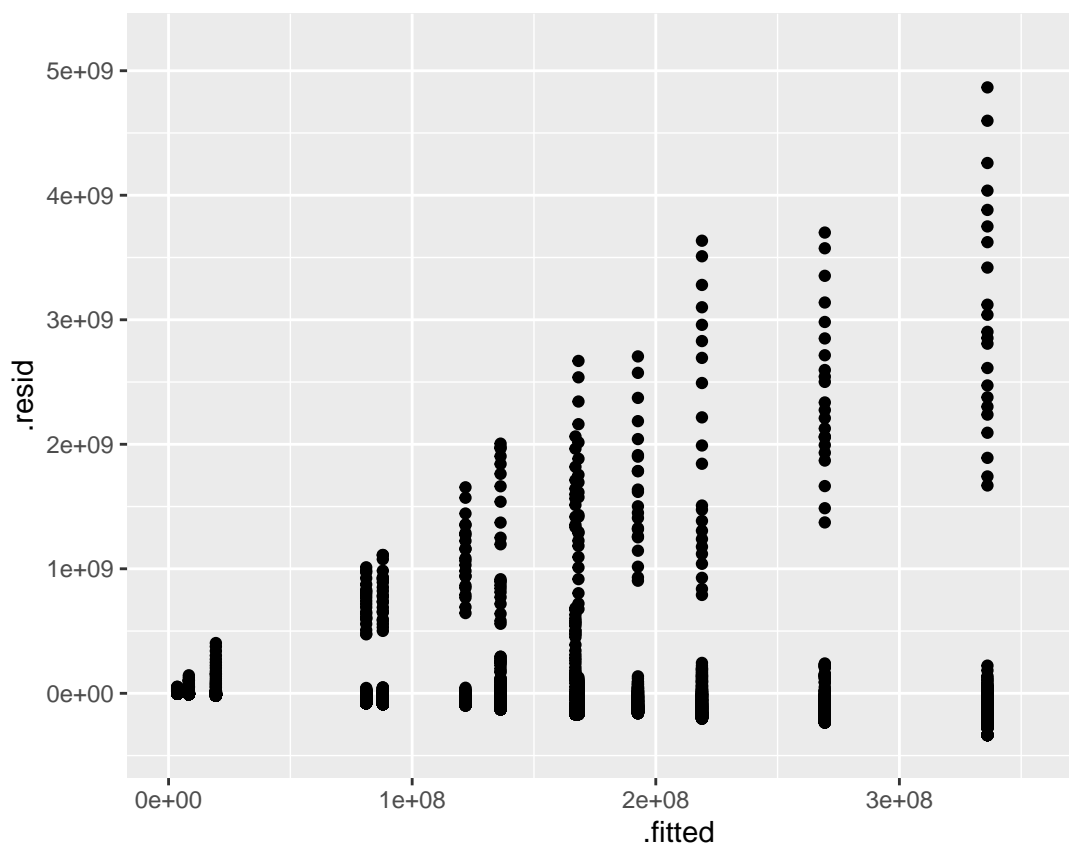
```
##           Df Sum Sq Mean Sq F value Pr(>F)
## agg_cause   14  19152   1368.0    521.9 <2e-16 ***
## Residuals 6365   16685     2.6
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Based on the p-value here of $<2e-16$, these data or more extreme data it is highly unlikely the null hypothesis is true. Therefore, we perform step-down tests using a Holm correction for multiple comparisons.

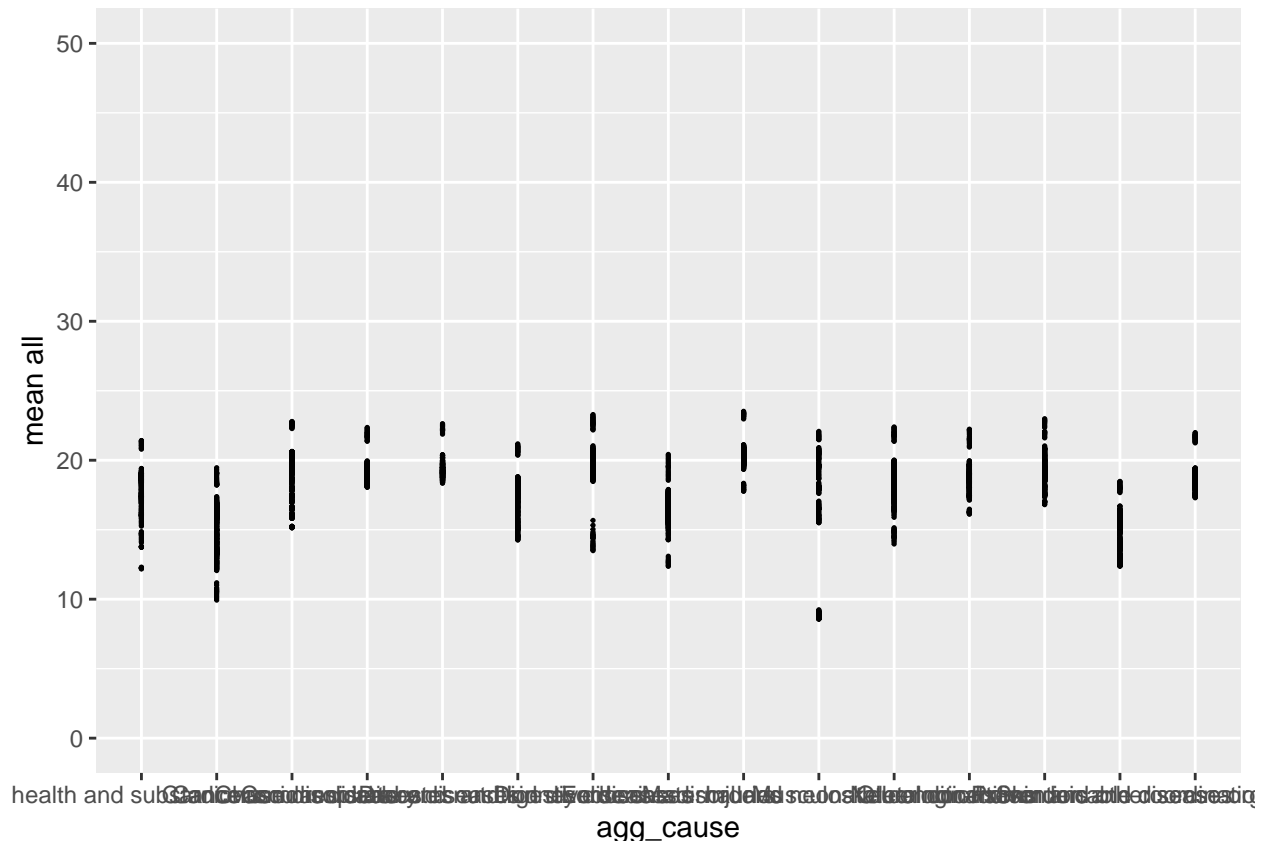
```
## [1] 92
```

The step-down t tests indicate 92 disease category pairs are different out of 105, indicating most disease categories do differ in the amount of government spending by the emergency department. !!not sure how to interpret anova i dont think this is right

```
## # A tibble: 15 x 5
##   term                                estimate std.error statistic  p.value
##   <chr>                                <dbl>     <dbl>     <dbl>    <dbl>
## 1 (Intercept)                        8.80e7 22716720.     3.87 1.09e- 4
## 2 agg_causeCancers                   -7.96e7 32126294.    -2.48 1.32e- 2
## 3 agg_causeCardiovascular diseases   3.16e8 32126294.    9.83 1.27e-22
## 4 agg_causeChronic respiratory diseases 1.05e8 32546263.    3.22 1.30e- 3
## 5 agg_causeCommunicable and nutrition d~ 1.81e8 32126294.    5.65 1.70e- 8
## 6 agg_causeDiabetes and kidney diseases -6.83e6 32126294.   -0.212 8.32e- 1
## 7 agg_causeDigestive diseases        2.48e8 32126294.    7.73 1.29e-14
## 8 agg_causeEndocrine disorders       -6.85e7 32126294.   -2.13 3.29e- 2
## 9 agg_causeInjuries                  3.27e8 32126294.   10.2 3.36e-24
## 10 agg_causeMaternal and neonatal condit~ 7.91e7 38135560.    2.07 3.82e- 2
## 11 agg_causeMusculoskeletal conditions  8.03e7 32126294.    2.50 1.25e- 2
## 12 agg_causeNeurological disorders    4.83e7 32126294.    1.50 1.33e- 1
## 13 agg_causeOther non-communicable disea~ 1.31e8 32126294.    4.08 4.61e- 5
## 14 agg_causePrevention and coordination -8.44e7 32126294.   -2.63 8.64e- 3
## 15 agg_causeSkin and other sense organ d~ 3.39e7 32126294.    1.06 2.91e- 1
```



!!this is weird, we need to discuss



Age

!! had to take out the observations with “All Ages” because I think it will just mess up the pairs but let me know what you think or whether you think there’s anything we can do with that group

We wonder whether there is a correlation between government healthcare expenditures in the emergency department and age. The age variable is categorical, split into 19 groups that generally include 5 years each, apart from the first (<1 year) and last (85 plus) groups.

To address this question, we began by using an overall test with ANOVA.

Below is an overall test of the null hypothesis that all of the means for age groups across the years are equal, as opposed to the alternative that at least one mean is different.

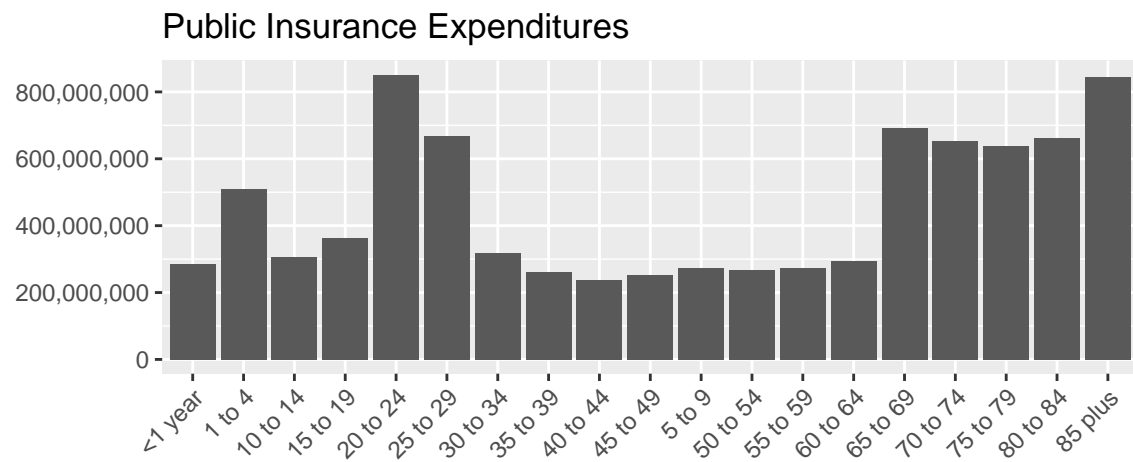
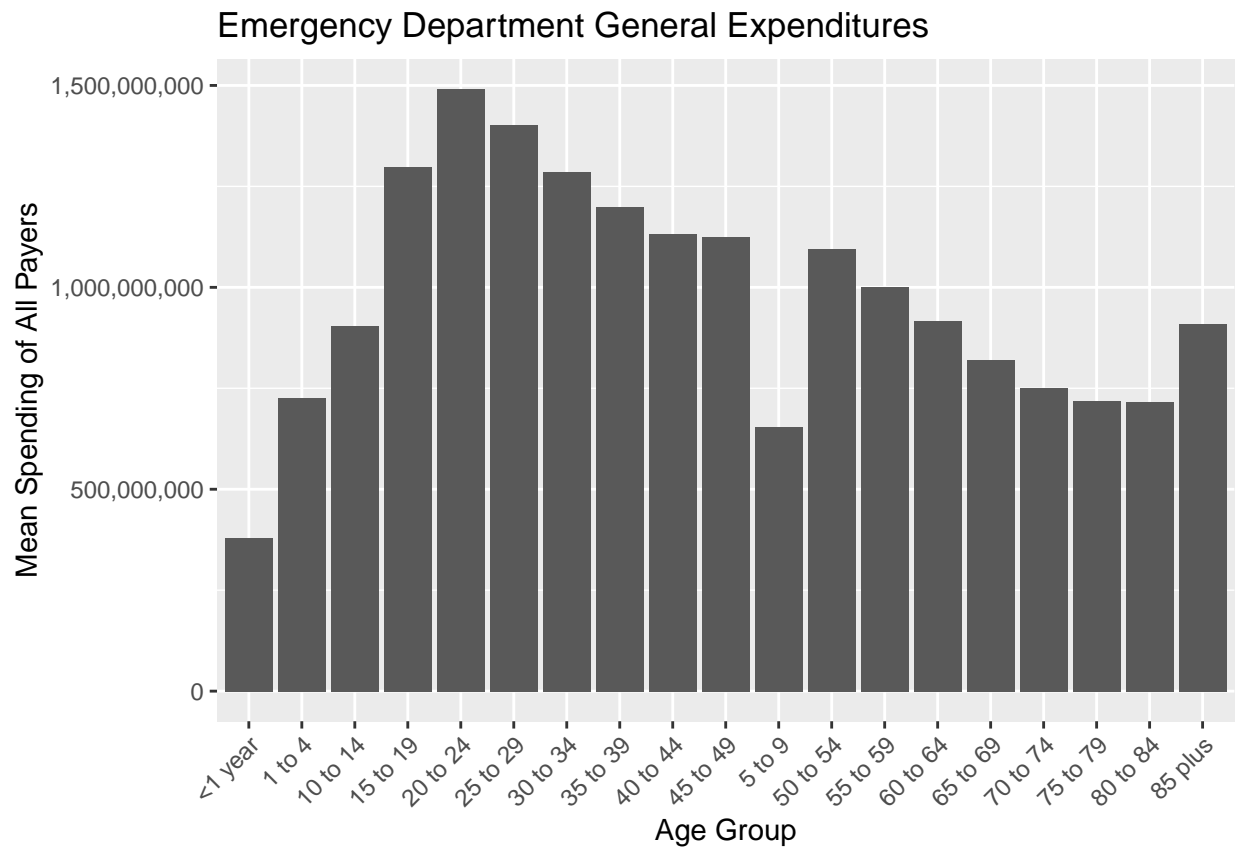
```
##           Df      Sum Sq   Mean Sq F value Pr(>F)
## age_group_name  18 2.843e+19 1.579e+18  29.45 <2e-16 ***
## Residuals      6031 3.235e+20 5.364e+16
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

In this F-test ($ndf = 18$, $ddf = 6229$), a significant difference among age groups was identified. Our p-value tells us that this data (or data more extreme) would be very unlikely if the null hypothesis were true because it shows statistical significance at an alpha well below 0.05. Therefore, we reject the null hypothesis that the mean expenditures for all age groups are equal.

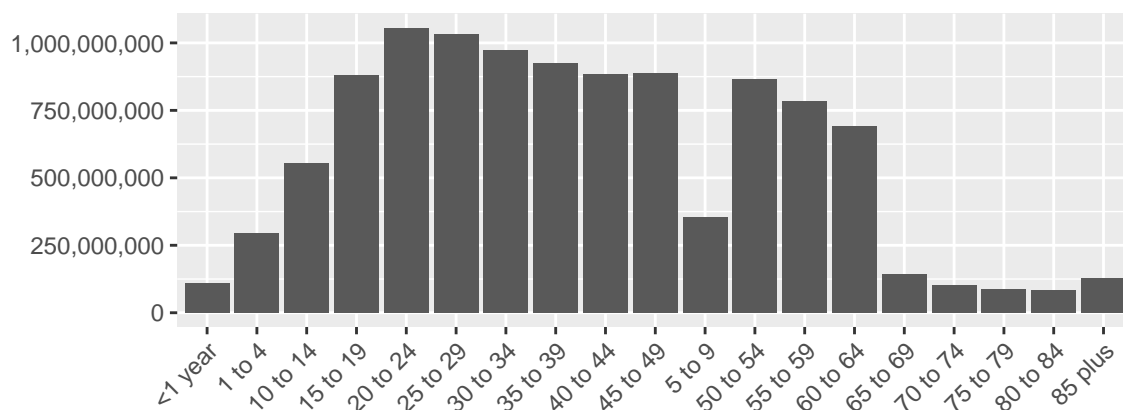
To see which specific means may be different from one another, we used planned step-down tests with a Holm correction to minimize Type I errors.

```
## [1] 98
```

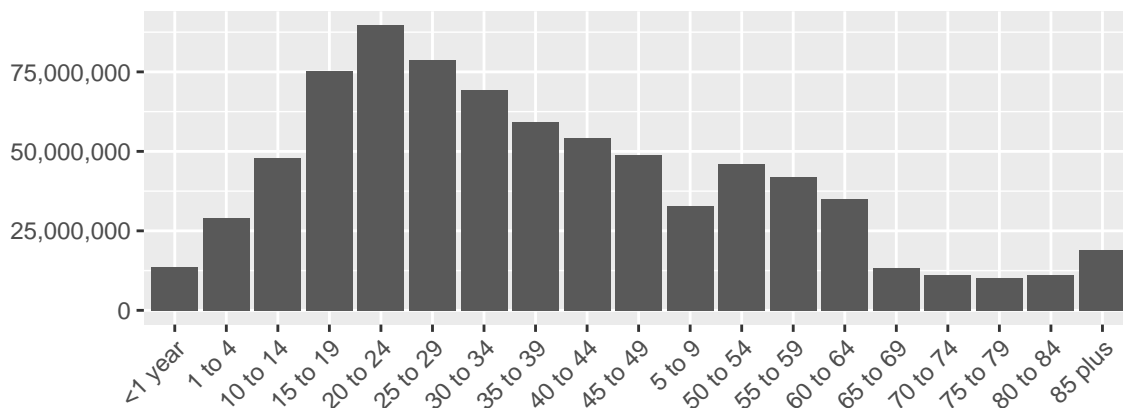
The pairwise t-tests used for our ANOVA step-down tests suggest that there are 97 different age pairs out of the 171 possible combinations. This tells us that more age pairs are different than are similar and that therefore the majority of age group pairs differ in terms of mean expenditures.



Private Insurance Expenditures



Out of Pocket Expenditures



Gender and Age Interaction

```
## # A tibble: 3 x 5
##   term          estimate std.error statistic  p.value
##   <chr>          <dbl>     <dbl>     <dbl>    <dbl>
## 1 (Intercept)    17.3      0.0445     390.      0
## 2 sexMale       -0.104     0.0574     -1.80  0.0712
## 3 age_group_id   0.00328   0.000887     3.70 0.000216
```

```
## [1] 0.002338971
```

```
## # A tibble: 4 x 5
##   term          estimate std.error statistic p.value
##   <chr>          <dbl>     <dbl>     <dbl>    <dbl>
## 1 (Intercept)    17.3      0.0483     359.      0
## 2 sexMale       -0.0803    0.0694     -1.16  0.247
## 3 age_group_id   0.00379   0.00123     3.08  0.00211
## 4 sexMale:age_group_id -0.00106  0.00177     -0.595 0.552
```

```
## [1] 0.002237904
```

```
## # A tibble: 3 x 5
##   term          estimate std.error statistic  p.value
##   <chr>          <dbl>     <dbl>     <dbl>    <dbl>
```

```
## 1 (Intercept)    17.1      0.0508    337.    0
## 2 sexMale        -0.0463    0.0656    -0.707 4.80e- 1
## 3 age_group_id   -0.0132    0.00101   -13.0   3.18e-38
## [1] 0.02563492

## # A tibble: 4 x 5
##   term                estimate std.error statistic  p.value
##   <chr>                <dbl>     <dbl>     <dbl>    <dbl>
## 1 (Intercept)         17.1      0.0552    310.     0
## 2 sexMale              0.00254    0.0793     0.0320 9.74e- 1
## 3 age_group_id        -0.0121    0.00141    -8.60   1.02e-17
## 4 sexMale:age_group_id -0.00222    0.00203    -1.10   2.73e- 1
## [1] 0.02566573

## # A tibble: 3 x 5
##   term                estimate std.error statistic  p.value
##   <chr>                <dbl>     <dbl>     <dbl>    <dbl>
## 1 (Intercept)         14.9      0.0485    308.     0
## 2 sexMale             -0.0613    0.0626    -0.980 3.27e- 1
## 3 age_group_id        -0.0113    0.000967   -11.7   3.10e-31
## [1] 0.02080634

## # A tibble: 4 x 5
##   term                estimate std.error statistic  p.value
##   <chr>                <dbl>     <dbl>     <dbl>    <dbl>
## 1 (Intercept)         14.9      0.0526    283.     0
## 2 sexMale             -0.0181    0.0757    -0.239 8.11e- 1
## 3 age_group_id        -0.0103    0.00134    -7.70   1.59e-14
## 4 sexMale:age_group_id -0.00197    0.00193    -1.02   3.09e- 1
## [1] 0.02081155
```

In order to test the possibility that there is a joint interaction of gender and age, a main effects and interaction effects linear regression model has been fit to the data. As a whole, it shows that the interaction of gender and age slightly increases the accuracy of the regression for public and private spending as seen by the increased adjusted R^2 value. However, for out-of-pocket spending, it decreases the adjusted R^2 value. Nevertheless, overall, the adjusted R^2 values for all three types of spending are incredibly low, which further point to our conclusion that age may not affect the level of spending from different sources.

Age and Disease Type Interaction

```
## # A tibble: 16 x 5
##   term                estimate std.error statistic  p.value
##   <chr>                <dbl>     <dbl>     <dbl>    <dbl>
## 1 (Intercept)         17.1      0.0780    219.     0
## 2 agg_causeCancers     -2.68     0.109    -24.6   3.75e-128
## 3 agg_causeCardiovascular diseases  1.16     0.109     10.7   1.92e- 26
## 4 agg_causeChronic respiratory diseases  1.27     0.110     11.5   2.51e- 30
## 5 agg_causeCommunicable and nutrition d~ 1.59     0.109     14.6   1.68e- 47
## 6 agg_causeDiabetes and kidney diseases  0.0502    0.109     0.462 6.44e- 1
## 7 agg_causeDigestive diseases  1.64     0.109     15.1   1.30e- 50
## 8 agg_causeEndocrine disorders  -1.27     0.109    -11.7   2.94e- 31
## 9 agg_causeInjuries      1.97     0.109     18.1   1.68e- 71
## 10 agg_causeMaternal and neonatal condit~ -3.14     0.129    -24.4   1.03e-125
## 11 agg_causeMusculoskeletal conditions  0.806     0.109     7.42   1.32e- 13
```

```
## 12 agg_causeNeurological disorders      0.760    0.109      7.00 2.77e- 12
## 13 agg_causeOther non-communicable disea~ 1.23     0.109     11.3 1.65e- 29
## 14 agg_causePrevention and coordination -3.06     0.109    -28.2 3.71e-165
## 15 agg_causeSkin and other sense organ d~ 0.863     0.109      7.95 2.22e- 15
## 16 age_group_id                        0.00334  0.000623    5.35 9.00e- 8
```

```
## [1] 0.506887
```

```
## # A tibble: 30 x 5
```

##	term	estimate	std.error	statistic	p.value
##	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	(Intercept)	17.3	0.0907	191.	0
## 2	agg_causeCancers	-3.14	0.128	-24.5	1.27e-126
## 3	agg_causeCardiovascular diseases	0.630	0.128	4.91	9.52e- 7
## 4	agg_causeChronic respiratory diseases	1.06	0.129	8.22	2.35e- 16
## 5	agg_causeCommunicable and nutrition d~	1.42	0.128	11.0	4.05e- 28
## 6	agg_causeDiabetes and kidney diseases	-0.244	0.128	-1.90	5.70e- 2
## 7	agg_causeDigestive diseases	1.48	0.128	11.6	1.19e- 30
## 8	agg_causeEndocrine disorders	-1.44	0.128	-11.2	6.78e- 29
## 9	agg_causeInjuries	1.72	0.128	13.4	2.25e- 40
## 10	agg_causeMaternal and neonatal condit~	-2.34	0.154	-15.2	4.33e- 51

```
## # ... with 20 more rows
```

```
## [1] 0.529351
```

```
## # A tibble: 16 x 5
```

##	term	estimate	std.error	statistic	p.value
##	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	(Intercept)	16.4	0.0903	182.	0
## 2	agg_causeCancers	-2.24	0.126	-17.8	1.48e- 69
## 3	agg_causeCardiovascular diseases	1.62	0.126	12.9	1.75e- 37
## 4	agg_causeChronic respiratory diseases	1.71	0.127	13.4	1.13e- 40
## 5	agg_causeCommunicable and nutrition d~	2.05	0.126	16.3	2.02e- 58
## 6	agg_causeDiabetes and kidney diseases	-0.245	0.126	-1.94	5.19e- 2
## 7	agg_causeDigestive diseases	2.43	0.126	19.3	6.39e- 81
## 8	agg_causeEndocrine disorders	-0.793	0.126	-6.31	2.99e- 10
## 9	agg_causeInjuries	3.44	0.126	27.3	9.02e-156
## 10	agg_causeMaternal and neonatal condit~	-3.16	0.149	-21.2	2.27e- 96
## 11	agg_causeMusculoskeletal conditions	1.50	0.126	11.9	2.36e- 32
## 12	agg_causeNeurological disorders	1.32	0.126	10.5	1.29e- 25
## 13	agg_causeOther non-communicable disea~	2.03	0.126	16.2	1.27e- 57
## 14	agg_causePrevention and coordination	-2.64	0.126	-21.0	2.18e- 94
## 15	agg_causeSkin and other sense organ d~	1.41	0.126	11.2	8.06e- 29
## 16	age_group_id	-0.0131	0.000722	-18.2	4.45e- 72

```
## [1] 0.5054947
```

```
## # A tibble: 30 x 5
```

##	term	estimate	std.error	statistic	p.value
##	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	(Intercept)	16.7	0.106	157.	0
## 2	agg_causeCancers	-2.77	0.151	-18.4	1.21e-73
## 3	agg_causeCardiovascular diseases	1.05	0.151	6.96	3.67e-12
## 4	agg_causeChronic respiratory diseases	1.50	0.152	9.90	6.18e-23
## 5	agg_causeCommunicable and nutrition di~	1.82	0.151	12.1	2.51e-33
## 6	agg_causeDiabetes and kidney diseases	-0.647	0.151	-4.30	1.74e- 5
## 7	agg_causeDigestive diseases	2.31	0.151	15.3	5.24e-52

```
## 8 agg_causeEndocrine disorders -1.04 0.151 -6.94 4.46e-12
## 9 agg_causeInjuries 3.23 0.151 21.4 1.91e-98
## 10 agg_causeMaternal and neonatal conditi~ -2.78 0.181 -15.4 2.21e-52
## # ... with 20 more rows
```

```
## [1] 0.5149051
```

```
## # A tibble: 16 x 5
```

##	term	estimate	std.error	statistic	p.value
##	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	(Intercept)	14.7	0.0850	173.	0
## 2	agg_causeCancers	-2.75	0.118	-23.3	6.48e-115
## 3	agg_causeCardiovascular diseases	0.975	0.118	8.23	2.17e-16
## 4	agg_causeChronic respiratory diseases	1.11	0.120	9.26	2.68e-20
## 5	agg_causeCommunicable and nutrition d~	1.67	0.118	14.1	1.02e-44
## 6	agg_causeDiabetes and kidney diseases	-0.428	0.118	-3.62	3.03e-4
## 7	agg_causeDigestive diseases	1.83	0.118	15.5	5.56e-53
## 8	agg_causeEndocrine disorders	-0.917	0.118	-7.75	1.08e-14
## 9	agg_causeInjuries	2.65	0.118	22.4	9.40e-107
## 10	agg_causeMaternal and neonatal conditi~	-3.52	0.141	-25.1	1.97e-132
## 11	agg_causeMusculoskeletal conditions	0.888	0.118	7.51	6.94e-14
## 12	agg_causeNeurological disorders	0.736	0.118	6.22	5.30e-10
## 13	agg_causeOther non-communicable disea~	1.57	0.118	13.3	1.30e-39
## 14	agg_causePrevention and coordination	-3.34	0.118	-28.2	1.73e-165
## 15	agg_causeSkin and other sense organ d~	0.841	0.118	7.10	1.37e-12
## 16	age_group_id	-0.0112	0.000680	-16.5	3.10e-60

```
## [1] 0.5161683
```

```
## # A tibble: 30 x 5
```

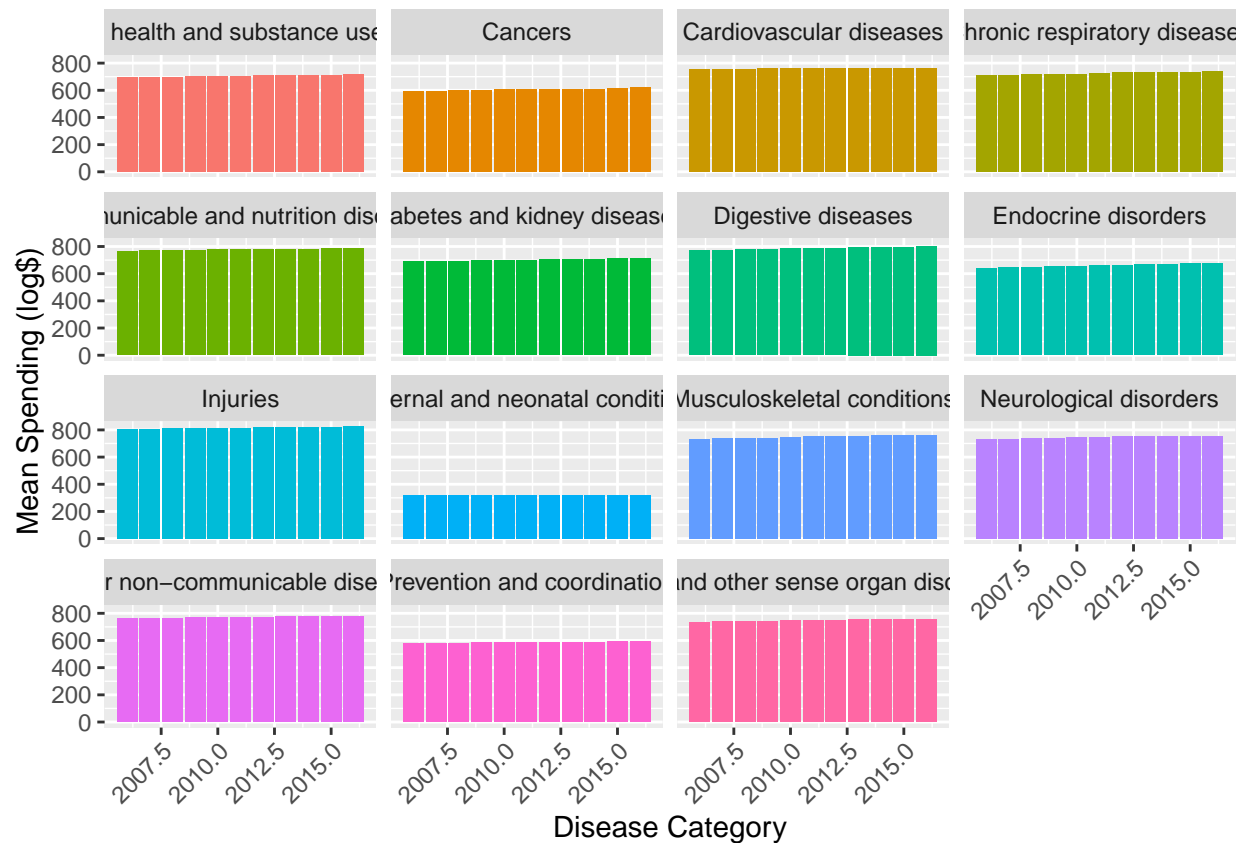
##	term	estimate	std.error	statistic	p.value
##	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	(Intercept)	14.9	0.100	149.	0
## 2	agg_causeCancers	-3.18	0.142	-22.4	2.10e-107
## 3	agg_causeCardiovascular diseases	0.404	0.142	2.85	4.34e-3
## 4	agg_causeChronic respiratory diseases	0.893	0.143	6.26	4.13e-10
## 5	agg_causeCommunicable and nutrition d~	1.50	0.142	10.6	5.92e-26
## 6	agg_causeDiabetes and kidney diseases	-0.797	0.142	-5.63	1.91e-8
## 7	agg_causeDigestive diseases	1.68	0.142	11.9	3.63e-32
## 8	agg_causeEndocrine disorders	-1.14	0.142	-8.08	7.94e-16
## 9	agg_causeInjuries	2.39	0.142	16.9	1.79e-62
## 10	agg_causeMaternal and neonatal conditi~	-3.09	0.170	-18.2	7.93e-72

```
## # ... with 20 more rows
```

```
## [1] 0.5261024
```

```
##Spending Over Time
```

!! I kinda like this but idk if it adds anything but it is fun, need to make the words smaller so you can read it



!! can we divide this to have a predictor for each year?

```
## # A tibble: 2 x 5
##   term      estimate std.error statistic    p.value
##   <chr>      <dbl>    <dbl>    <dbl>    <dbl>
## 1 (Intercept) -92.1      18.8      -4.89 0.00000102
## 2 year_id      0.0548    0.00936     5.85 0.00000000516
```