

## 1. Descriptive analysis for qualitative variables

### Research Question:

How do Alabama hospitals score in national comparison measure for readmission rates?

### Summary:

Most of Alabama's hospitals in this sample fall into the Same as National Average category (66.25%) followed by Above National Average (18.75%) and Below National Average (15%).

### R Results:

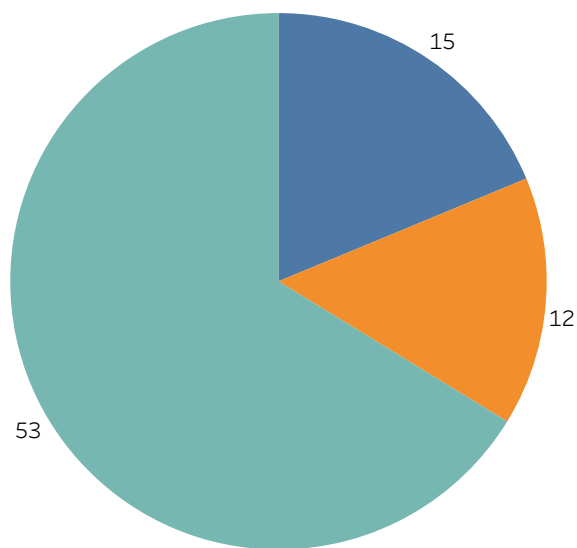
```
> local({
+   .Table <- with(Alabama, table(Readmission_1))
+   cat("\ncounts:\n")
+   print(.Table)
+   cat("\npercentages:\n")
+   print(round(100*.Table/sum(.Table), 2))
+ })

counts:
Readmission_1
Below the National average Same as the National average Above the National average
               12              53              15

percentages:
Readmission_1
Below the National average Same as the National average Above the National average
               15.00              66.25              18.75
```

### Alabama Hospital Readmission Comparisons

■ Above the National average  
■ Below the National average  
■ Same as the National average



## 2. Descriptive analysis for quantitative variables

### Research Question:

How do Alabama hospitals score for overall ratings by CBSA type?

### Summary:

When comparing the means for overall ratings in Alabama hospitals, Rural had the highest with an average score of 3.35, then Micro with 3.07, then Metro with 2.91.

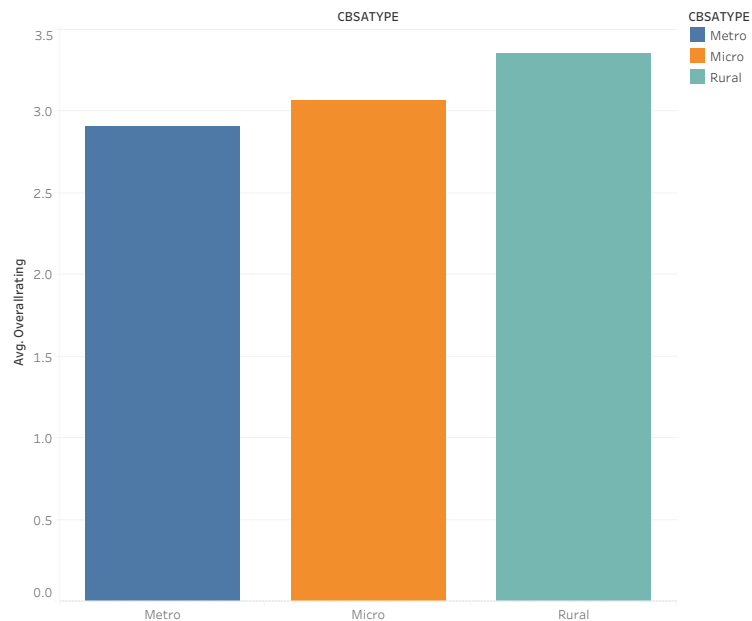
### R Results:

```
> numSummary(Alabama[, "Overallrating", drop=FALSE], groups=Alabama$CBSATYPE,
+ statistics=c("mean", "sd", "se(mean)", "IQR", "quantiles", "cv"), quantiles=c(0,.25,.5,
+ .75,1))
```

	mean	sd	se(mean)	IQR	cv	0%	25%	50%	75%	100%	Overallrating:n
Metro	2.906977	0.7500461	0.1143810	1 0.2580159	2	2	3	3	5		43
Micro	3.066667	0.4577377	0.1181874	0 0.1492623	2	3	3	3	4		15
Rural	3.350000	0.7451598	0.1666228	1 0.2224358	2	3	3	4	5		20

```
Overallrating:NA
Metro      4
Micro      0
Rural      5
```

Average Overall Rating by CBSA Type in Alabama



## 2. Single Sample t-test

### Research question:

Is there a difference in the overall rating mean for critical access hospitals from the overall population mean?

### Null Hypothesis:

Critical Access Hospitals do not differ from the overall population (mean= 3.059) with respect to the overall rating.

### Summary:

A single-sample t-test was conducted to test the null hypothesis that the average overall rating of critical access hospitals was not different from the overall population mean of 3.059. There was a statistically significant difference ( $p < 0.05$ ) in means between critical access hospitals ( $\bar{x} = 3.323$ ) and population ( $\mu = 3.059$ );  $t(596) = 11.479$ . The effect size, Cohen's  $d = 0.469$  indicating 0.469 standard deviation above the population mean.

### R Results:

```
Overallrating
Min.      :1.000
1st Qu.:3.000
Median   :3.000
Mean     :3.059
3rd Qu.:4.000
Max.     :5.000
NA's     :2421
```

```

HospitalType_1
Acute Care Hospitals :3370
Childrens            : 99
Critical Access Hospitals:1338
NA's                 :1198
```

### One Sample t-test

```
data: Overallrating
t = 11.479, df = 596, p-value < 2.2e-16
alternative hypothesis: true mean is not equal to 3.059
95 percent confidence interval:
 3.278066 3.368500
sample estimates:
mean of x
 3.323283
```

## 4. Independent Samples t-test

### Research Question:

Does ER status affect excessive readmission ratio for AMI?

### Null Hypothesis:

ER status does not affect excessive readmission ratio for AMI.

### Summary:

Because the significance (probability) level is greater than .05 ( $p < F = .1492$ ), we fail to reject the null hypothesis that there is no difference in average/mean excessive readmission ratio for AMI between providers with an ER and without. We conclude that the mean excessive readmission ratio for AMI of patients with an ER are statistically the same as those without an ER.

An independent-samples t-test was conducted to compare the average excessive readmission ratio for AMI for providers with and without an ER. There was no statistically significant difference in the excessive readmission ratio for AMI between providers without an ER (mean=0.994, SD=0.058) and with an ER (mean=1.002, SD=0.067).

### R Results:

```
> leveneTest(ExcessReadRatio_AMI ~ ER_1, data=Dataset, center="mean")
Levene's Test for Homogeneity of Variance (center = "mean")
      Df F value Pr(>F)
group  1  2.0815 0.1492
      2117

> numSummary(Dataset[, "ExcessReadRatio_AMI", drop=FALSE], groups=Dataset$ER_1,
+   statistics=c("mean", "sd", "se(mean)", "IQR", "quantiles"), quantiles=c(0,.25,.5,.75,.100),
+   mean      sd      se(mean)  IQR    0%    25%    50%    75%    100%
No  0.9944594 0.05818135 0.010285107 0.0412 0.8359 0.98415 1.01115 1.02535 1.0903
Yes 1.0023157 0.06655378 0.001456839 0.0825 0.7043 0.96045 1.00020 1.04295 1.2491
      ExcessReadRatio_AMI:n ExcessReadRatio_AMI:NA
No                          32                      263
Yes                         2087                     2425

      Two Sample t-test

data: ExcessReadRatio_AMI by ER_1
t = -0.66385, df = 2117, p-value = 0.5069
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -0.03106486  0.01535218
sample estimates:
mean in group No mean in group Yes
      0.9944594      1.0023157
```

## 5. One Way ANOVA

### Research Question:

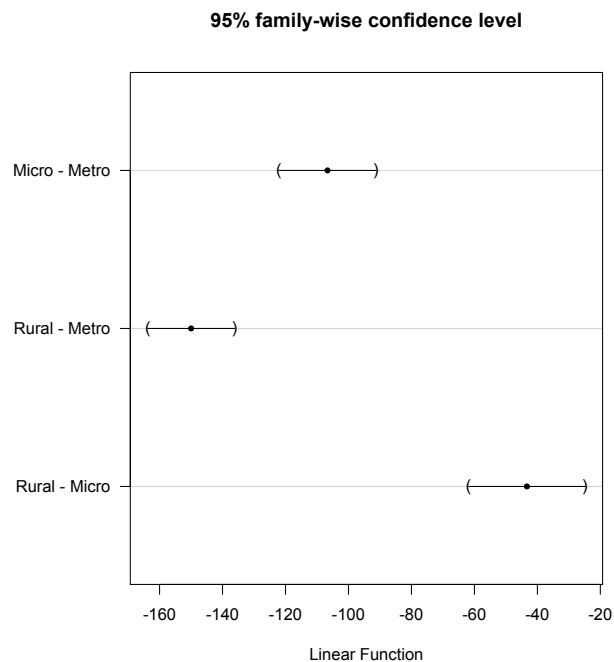
Is there a difference in number of beds across different CBSA types?

### Null Hypothesis:

CBSA type does not have a statistically significant effect on number of beds.

### Summary:

A one-way between-groups analysis of variance was conducted to explore the impact of CBSA category on number of beds. There was a statistically significant difference at the 0.001 level in number of beds for three CBSA categories [ $F(2, 5918) = 386.9, p = .000$ ]. The effect size, calculated using Eta-squared, was 11.56%. Post-hoc comparisons using the Tukey Contrasts indicated that the mean number of beds was significantly different between Micro and Metro, Rural and Metro, and Rural and Micro. Metro ( $M = 193.5, SD = 210.4$ ), Micro ( $M = 86.8, SD = 80.0$ ), and Rural ( $M = 43.5, SD = 43.4$ )



**R Results:**

```
> summary(AnovaModel.1)
              Df    Sum Sq Mean Sq F value Pr(>F)
CBSATYPE      2  23720396 11860198   386.9 <2e-16 ***
Residuals    5918 181428554   30657
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
84 observations deleted due to missingness

> with(Dataset, numSummary(HOSPBD, groups=CBSATYPE, statistics=c("mean", "sd")))
      mean      sd data:n
Metro 193.46780 210.35901  3929
Micro  86.80980  79.96825   857
Rural  43.48194  43.40969  1135
```

## Simultaneous Tests for General Linear Hypotheses

## Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = HOSPBD ~ CBSATYPE, data = Dataset)
```

## Linear Hypotheses:

	Estimate	Std. Error	t value	Pr(> t )
Micro - Metro == 0	-106.658	6.601	-16.157	< 0.0000001 ***
Rural - Metro == 0	-149.986	5.900	-25.420	< 0.0000001 ***
Rural - Micro == 0	-43.328	7.924	-5.468	0.000000111 ***

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Adjusted p values reported -- single-step method)
```

## Simultaneous Confidence Intervals

## Multiple Comparisons of Means: Tukey Contrasts

```
Fit: aov(formula = HOSPBD ~ CBSATYPE, data = Dataset)
```

```
Quantile = 2.3327
```

```
95% family-wise confidence level
```

## Linear Hypotheses:

	Estimate	lwr	upr
Micro - Metro == 0	-106.6580	-122.0566	-91.2594
Rural - Metro == 0	-149.9859	-163.7495	-136.2222
Rural - Micro == 0	-43.3279	-61.8113	-24.8444

```
Metro Micro Rural
  "c"   "b"   "a"
```

## 6. Chi-Square test

### Research Question:

Does hospital type have an effect on CBSA type?

### Null Hypothesis:

CBSA type is unrelated to hospital type.

### Summary:

A Chi-square test was conducted to explore whether CBSA type is related to hospital type. A p-value of 0.000 shows there is a statistically significant relationship between CBSA type and hospital type. For example, most acute care hospitals (75.3%) were CBSA type metro, all (100%) of children's hospitals were also metro. Most (63.2%) critical access hospitals were rural and a relatively small percentage (8.3%) of acute care hospitals were rural.

### R Results:

```
Frequency table:
      HospitalType_1
CBSATYPE Acute Care Hospitals Childrens Critical Access Hospitals
Metro      2493           94              257
Micro       543            0              229
Rural       274            0              833

Column percentages:
      HospitalType_1
CBSATYPE Acute Care Hospitals Childrens Critical Access Hospitals
Metro      75.3           100             19.5
Micro      16.4            0             17.4
Rural       8.3            0             63.2
Total     100.0           100            100.1
Count     3310.0          94            1319.0

Pearson's Chi-squared test

data: .Table
X-squared = 1763.9, df = 4, p-value < 2.2e-16
```

## 7. Correlation test

### Research Question:

Is a high discharge count for AMI associated with excessive readmission ratios?

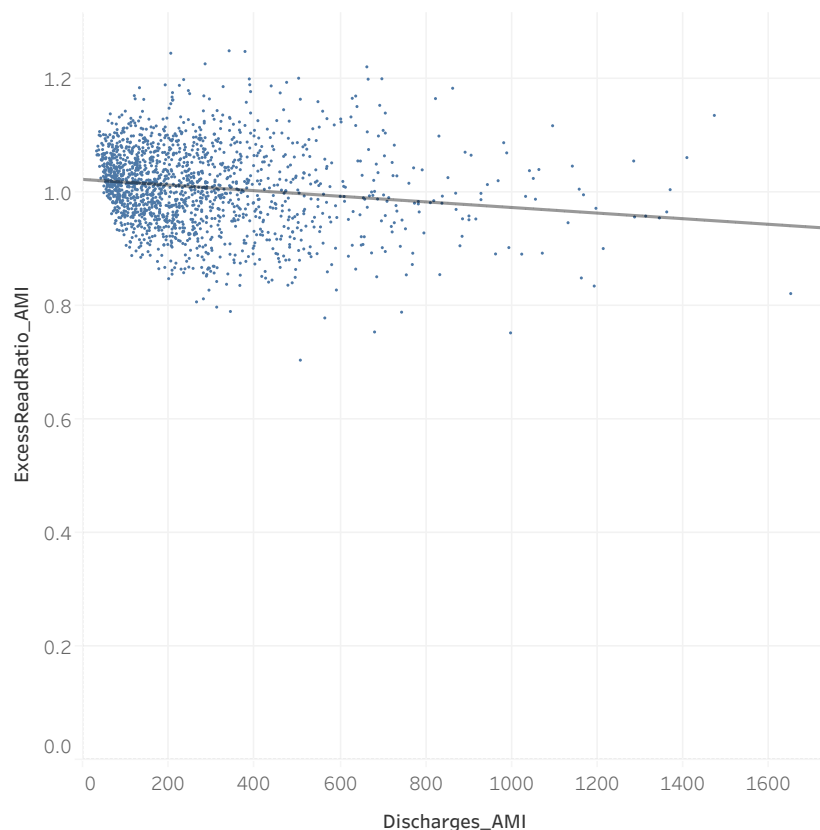
### Null Hypothesis:

There is no statistically significant relationship between readmissions for AMI and excessive readmission ratio for AMI.

### Summary:

The relationship between beginning number of discharges and excessive readmissions ratios for AMI was investigated using Pearson product-moment correlation test. Preliminary analyses were performed to ensure no violation of assumptions of normality and linearity. A statistically significant negative correlation between the two variables was observed [ $r = -0.197$   $n = 1682$ ,  $p < 0.001$ ], indicating that high values for discharges are associated with lower excessive readmission ratios.

Excess Readmission Ratio vs Number of Discharges for AMI





**R Results:**

```
Pearson's product-moment correlation  
data: Discharges_AMI and ExcessReadRatio_AMI  
t = -6.2606, df = 1680, p-value = 4.859e-10  
alternative hypothesis: true correlation is not equal to 0  
95 percent confidence interval:  
-0.1973627 -0.1039452  
sample estimates:  
cor  
-0.150991
```

## 8. Correlation Matrix

### Research Question:

Is there a relationship between number of AMI discharges, AMI excessive readmissions ratio (ERR), and overall hospital rating?

### Null Hypothesis:

There is no statistically significant relationship between number of AMI discharges, AMI excessive readmissions ratio, and overall hospital rating.

### Summary:

The relationship between number of AMI discharges, AMI excessive readmissions ratio (ERR), and overall hospital rating (OHR) was investigated using Pearson product-moment correlation matrix. Preliminary analyses were performed to ensure no violation of assumptions of normality and linearity. A statistically significant negative correlation was observed between discharges and ERR [ $r = -0.151$ ,  $n = 1682$ ,  $p < 0.001$ ], and between ERR and OHR [ $r = -0.341$ ,  $n = 2119$ ,  $p < 0.001$ ], indicating that high ERR values are associated with lower OHR and discharge values. A statistically significant positive correlation was also observed between OHR and discharges [ $r = 0.0784$ ,  $n = 1708$ ,  $p < 0.05$ ], indicating that high values for OHR are associated with high discharge values.

Pearson Product-Moment Correlations among Hospital Performance Measures:

Measures	1. AMI Discharges	2. AMI ERR
1. AMI Discharges		
2. AMI ERR	-0.151**	
3. Overall Rating	0.0784*	-0.341**

\*\* $p < 0.001$ , \*  $p < 0.05$

Excess Readmission Ratio vs Number of Discharges for AMI

**R Results:**

## Pearson correlations:

	Discharges_AMI	ExcessReadRatio_AMI	Overallrating
Discharges_AMI	1.0000	-0.1510	0.0784
ExcessReadRatio_AMI	-0.1510	1.0000	-0.3414
Overallrating	0.0784	-0.3414	1.0000

## Number of observations:

	Discharges_AMI	ExcessReadRatio_AMI	Overallrating
Discharges_AMI	1726	1682	1708
ExcessReadRatio_AMI	1682	2119	2119
Overallrating	1708	2119	3584

## Pairwise two-sided p-values:

	Discharges_AMI	ExcessReadRatio_AMI	Overallrating
Discharges_AMI		<.0001	0.0012
ExcessReadRatio_AMI	<.0001		<.0001
Overallrating	0.0012	<.0001	

## Adjusted p-values (Holm's method)

	Discharges_AMI	ExcessReadRatio_AMI	Overallrating
Discharges_AMI		<.0001	0.0012
ExcessReadRatio_AMI	<.0001		<.0001
Overallrating	0.0012	<.0001	

## 9. Simple linear regression using a quantitative independent variable

### Research Question:

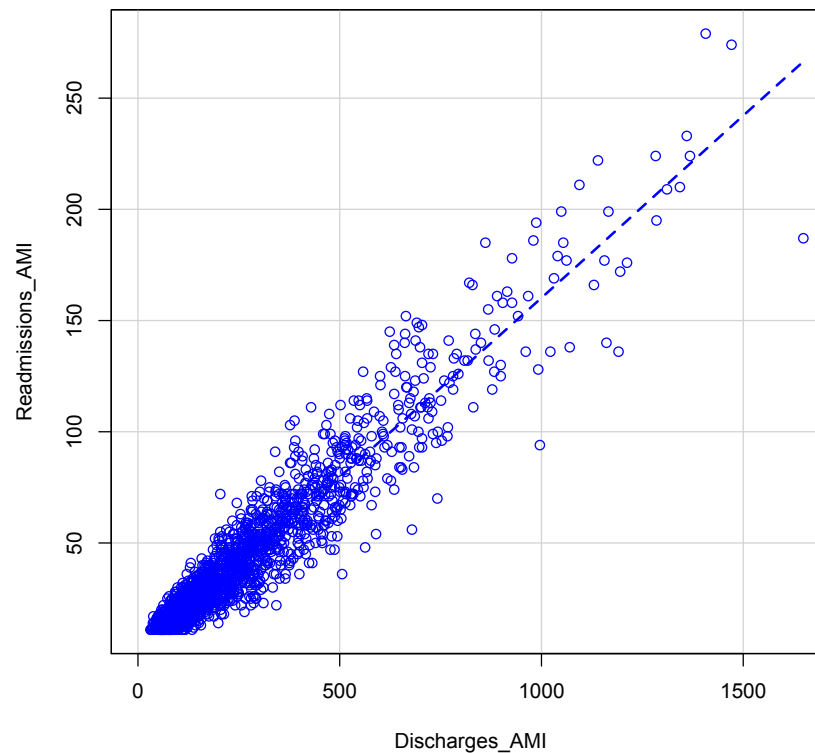
Can the number of readmissions be predicted by number of discharges in COPD patients?

### Null Hypothesis:

Number of COPD discharges does not predict stroke readmissions.

### Summary:

A simple regression analysis was conducted to test if AMI discharges significantly predicted readmissions. The results of the regression indicated AMI discharges explained 95.2% of variances of readmissions ( $R^2=.952$ ,  $F(1, 2568)= 5.075e+04$ ,  $p<.0001$ ).



**R Results:**

```

Call:
lm(formula = Readmissions_COPD ~ Discharges_COPD, data = Dataset)

Residuals:
    Min       1Q   Median       3Q      Max
-53.894  -5.696   0.213   5.747  87.479

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)   -4.321913   0.3691760  -11.71  <2e-16 ***
Discharges_COPD 0.2148850  0.0009539   225.28  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 11.16 on 2568 degrees of freedom
(3435 observations deleted due to missingness)
Multiple R-squared:  0.9518, Adjusted R-squared:  0.9518
F-statistic: 5.075e+04 on 1 and 2568 DF,  p-value: < 2.2e-16

```

## 10. Simple linear regression using a qualitative independent variable

### Research Question:

Does having an ER influence overall ratings?

### Null Hypothesis:

ER status does not affect overall ratings.

### Summary:

A simple regression analysis was conducted to test if ER status significantly predicted overall rating. The results of the regression indicated the ER status explained 0.22% of variances of current salary ( $R^2=0.0022$ ,  $F(1, 3582)=7.94$ ,  $p<0.01$ ). Overall rating for hospitals with an ER was .27 less than hospitals without an ER ( $p<0.01$ ).

### R Results:

```
Call:
lm(formula = Overallrating ~ ER, data = Dataset)

Residuals:
    Min       1Q   Median       3Q      Max
-2.32000 -0.05301 -0.05301  0.94699  1.94699

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)   3.32000    0.09378  35.404  < 2e-16 ***
ER            -0.26699    0.09477  -2.817  0.00487 **
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.8121 on 3582 degrees of freedom
(2421 observations deleted due to missingness)
Multiple R-squared:  0.002211, Adjusted R-squared:  0.001932
F-statistic: 7.937 on 1 and 3582 DF,  p-value: 0.004871
```