# **OLS** Regression

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## 10/23/2018

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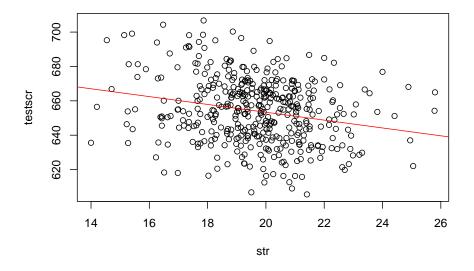
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## 1 Plot

#### 1.1 Scatter Plot

• Draw a scatter plot of the variable testscr against str:

```
attach(cars_data_small)
plot(str, testscr)
abline(lm(testscr ~ str , data = cars_data_small),col = "red")
```

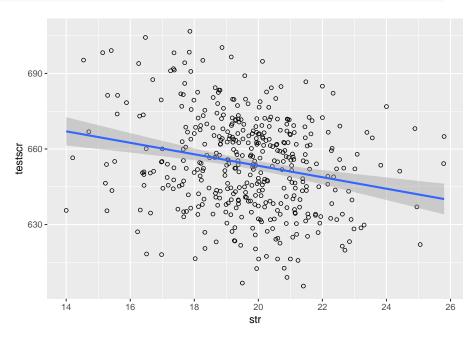


```
lm(formula, data, subset, weights, na.action,
  method = "qr", model = TRUE, x = FALSE, y = FALSE, qr = TRUE,
  singular.ok = TRUE, contrasts = NULL, offset, ...)
```

### 1.2 ggplot2

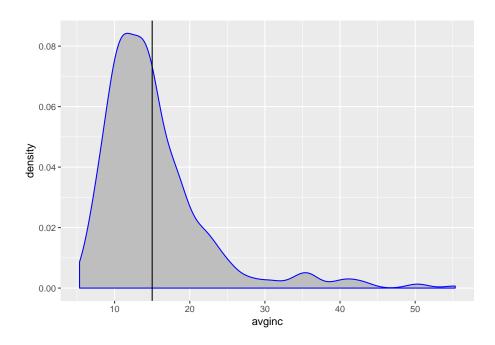
```
library("ggplot2")
ggplot(data =cars_data_small,aes(x=str, y=testscr)) +
```

```
geom_point(shape=1) + # Use hollow circles
geom_smooth(method=lm) # Add linear regression line
```

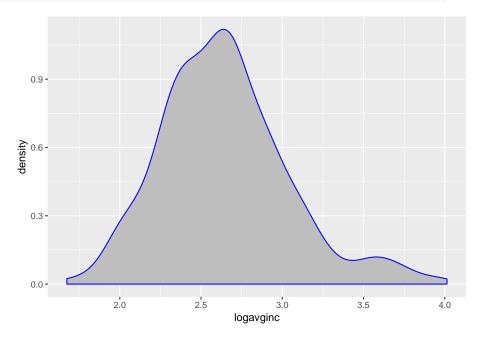


### 1.3 A kdensity distribution of income

```
cars_data$inc <- with(cars_data,avginc >=15)
ggplot(cars_data,aes(x=avginc))+
geom_density(fill="grey",color ="blue")+
geom_vline(xintercept = 15)
```

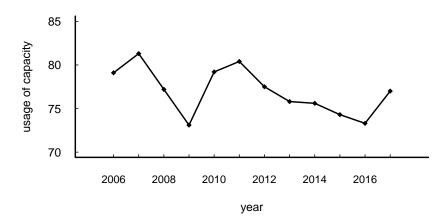


```
cars_data$logavginc <- log(cars_data$avginc)
ggplot(cars_data,aes(x=logavginc))+
geom_density(fill="grey",color ="blue")</pre>
```



#### 1.4 extra image

```
library(readxl)
image <- read_excel("/Users/admin/Desktop/teaching assistant/Econometrics/teaching assistant/Econometrics
```



2 T-TEST IN R 6

#### par(opar)

```
plot symbols: pch=

□ 0 ♦ 5 ⊕ 10 ■ 15 • 20 ▽ 25

○ 1 ▽ 6 ፟ 11 • 16 ○ 21

△ 2 ☒ 7 ⊞ 12 ▲ 17 □ 22

+ 3 ※ 8 ☒ 13 • 18 ♦ 23

× 4 ⊕ 9 ☒ 14 • 19 △ 24
```

#### 2 T-test in R

#### 2.1 single sample

• t-test for scores

```
summary(cars_data_small$testscr)
##
     Min. 1st Qu. Median
                              Mean 3rd Qu.
                                              Max.
##
     605.5
             640.0
                     654.5
                             654.2
                                     666.7
                                             706.8
t.test(cars_data_small$testscr,alternative = "two.sided",mu = 650)
##
    One Sample t-test
##
##
## data: cars_data_small$testscr
## t = 4.4708, df = 419, p-value = 1.005e-05
## alternative hypothesis: true mean is not equal to 650
## 95 percent confidence interval:
```

```
## 652.3291 655.9840
## sample estimates:
## mean of x
## 654.1565
```

## 3 OLS Regression

• https://cran.r-project.org/doc/contrib/Ricci-refcard-regression.pdf

```
fm1 <- lm(testscr ~ str,data = cars_data_small)</pre>
summary(fm1)
##
## Call:
## lm(formula = testscr ~ str, data = cars_data_small)
##
## Residuals:
      Min
               1Q Median
                               3Q
                                      Max
## -47.727 -14.251 0.483 12.822 48.540
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 698.9330
                           9.4675 73.825 < 2e-16 ***
               -2.2798
                           0.4798 -4.751 2.78e-06 ***
## str
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 18.58 on 418 degrees of freedom
## Multiple R-squared: 0.05124, Adjusted R-squared: 0.04897
## F-statistic: 22.58 on 1 and 418 DF, p-value: 2.783e-06
```

#### 3.1 OLS Regression 2

```
fm2 <- lm(testscr ~ str,data = cars_data)</pre>
summary(fm2)
##
## Call:
## lm(formula = testscr ~ str, data = cars_data)
##
## Residuals:
      Min
##
               1Q Median
                               3Q
                                      Max
## -47.727 -14.251 0.483 12.822 48.540
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 698.9330
                          9.4675 73.825 < 2e-16 ***
                          0.4798 -4.751 2.78e-06 ***
               -2.2798
## str
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 18.58 on 418 degrees of freedom
## Multiple R-squared: 0.05124,
                                  Adjusted R-squared: 0.04897
## F-statistic: 22.58 on 1 and 418 DF, p-value: 2.783e-06
    OLS Regression 3
fm3 <- lm(testscr ~ str+meal_pct,data = cars_data)</pre>
summary(fm3)
##
## Call:
## lm(formula = testscr ~ str + meal_pct, data = cars_data)
##
```

## str

```
## Residuals:
      Min
               1Q Median
                               ЗQ
                                      Max
## -33.657 -5.518 -0.168 5.645 34.494
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 702.91131
                          4.70025 149.548 < 2e-16 ***
## str
               -1.11723
                           0.24036 -4.648 4.5e-06 ***
## meal_pct
               -0.59975
                           0.01676 -35.775 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 9.222 on 417 degrees of freedom
## Multiple R-squared: 0.7668, Adjusted R-squared: 0.7657
## F-statistic: 685.8 on 2 and 417 DF, p-value: < 2.2e-16
3.3 OLS Regression 4
fm4 <- lm(testscr ~ str+meal_pct+computer,data = cars_data)</pre>
summary(fm4)
##
## Call:
## lm(formula = testscr ~ str + meal_pct + computer, data = cars_data)
## Residuals:
      Min
               1Q Median
                               ЗQ
##
                                      Max
## -33.638 -5.620 -0.144 5.690 34.544
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 7.031e+02 4.775e+00 147.239 < 2e-16 ***
```

-1.129e+00 2.471e-01 -4.566 6.55e-06 \*\*\*

```
## meal_pct    -5.999e-01  1.679e-02 -35.723 < 2e-16 ***
## computer    2.108e-04  1.052e-03  0.200  0.841
## ---
## Signif. codes:    0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.233 on 416 degrees of freedom
## Multiple R-squared: 0.7669, Adjusted R-squared: 0.7652
## F-statistic: 456.1 on 3 and 416 DF, p-value: < 2.2e-16</pre>
```

#### 3.4 OLS Regression 5

```
fm5 <- lm(testscr ~ str+I(str^2),data = cars_data)</pre>
summary(fm5)
##
## Call:
## lm(formula = testscr ~ str + I(str^2), data = cars_data)
##
## Residuals:
      Min
               1Q Median
                               3Q
                                      Max
## -47.349 -14.226
                   0.556 12.981 48.593
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 739.4456
                          60.7179 12.178
                                          <2e-16 ***
## str
               -6.4490
                           6.1906 -1.042
                                             0.298
## I(str^2)
              0.1063
                           0.1573 0.676
                                             0.500
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 18.59 on 417 degrees of freedom
## Multiple R-squared: 0.05228, Adjusted R-squared: 0.04773
## F-statistic: 11.5 on 2 and 417 DF, p-value: 1.374e-05
```

#### 3.5 OLS Regression 6

```
fm6 <- lm(testscr ~ str+str:el_pct,data = cars_data)</pre>
summary(fm6)
##
## Call:
## lm(formula = testscr ~ str + str:el_pct, data = cars_data)
##
## Residuals:
##
      Min
               1Q Median
                               ЗQ
                                      Max
## -49.053 -10.119 -0.704 9.873 43.898
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 677.610055 7.504411 90.295
                                              <2e-16 ***
## str
               -0.673273   0.387181   -1.739   0.0828 .
## str:el_pct -0.032357 0.001973 -16.398 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 14.51 on 417 degrees of freedom
## Multiple R-squared: 0.4232, Adjusted R-squared: 0.4204
## F-statistic: 153 on 2 and 417 DF, p-value: < 2.2e-16
```

符号	用 途
~	分隔符号,左边为响应变量,右边为解释变量。例如,要通过x、z和w预测y,代码为y ~ x + z + w
+	分隔预测变量
:	表示预测变量的交互项。例如,要通过 $x$ 、 $z$ $D_x$ 与 $z$ 的交互项预测 $y$ ,代码为 $y$ $\sim$ $x$ + $z$ + $x$ : $z$
*	表示所有可能交互项的简洁方式。代码y~ x * z * w可展开为y ~ x + z + w + x:z + x:w + z:w + x:z:w
^	表示交互项达到某个次数。代码y ~ (x + z + w)^2可展开为y ~ x + z + w + x:z + x:w + z:w
	表示包含除因变量外的所有变量。例如,若一个数据框包含变量 $x$ 、 $y$ 、 $z$ 和 $w$ ,代码 $y$ ~ . 可展开为 $y$ ~ $x$ + $z$ + $w$
-	减号,表示从等式中移除某个变量。例如,y~(x+z+w)^2-x:w可展开为y~x+z+w+x:z+z:w
-1	删除截距项。例如,表达式y ~ x - 1拟合y在x上的回归,并强制直线通过原点
I()	从算术的角度来解释括号中的元素。例如, $y\sim x+(z+w)^2$ 将展开为 $y\sim x+z+w+z$ :w。相反,代码 $y\sim x+z(z+w)^2$ )将展开为 $y\sim x+h$ ,h是一个由 $z$ 和w的平方和创建的新变量
function	可以在表达式中用的数学函数。例如,log(y)~x+z+w表示通过x、z和w来预测log(y)

## 4 Published Tables: Stargazer

```
#install.packages("stargazer",repos = "http://mirrors.ustc.edu.cn/CRAN/")
library("stargazer")
```

#### ##

## Please cite as:

- ## Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary Statistics T
- ## R package version 5.2.2. https://CRAN.R-project.org/package=stargazer

#### 4.1 Published Tables: Stargazer1

```
stargazer(fm1,fm3,fm4,title="Regression Results1")
```

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu % Date and time: 周一, 10 月 22, 2018 - 23:35:23

表 1: Regression Results1

	Dependent variable:			
	testscr			
	(1)	(2)	(3)	
str	$-2.280^{***}$	-1.117***	-1.129***	
	(0.480)	(0.240)	(0.247)	
meal_pct		-0.600***	$-0.600^{***}$	
<b>.</b>		(0.017)	(0.017)	
computer			0.0002	
•			(0.001)	
Constant	698.933***	702.911***	703.074***	
	(9.467)	(4.700)	(4.775)	
Observations	420	420	420	
$\mathbb{R}^2$	0.051	0.767	0.767	
Adjusted $\mathbb{R}^2$	0.049	0.766	0.765	
Residual Std. Error	18.581 (df = 418)	9.222 (df = 417)	9.233 (df = 416)	
F Statistic	$22.575^{***} (df = 1; 418)$	$685.756^{***} (df = 2; 417)$	$456.132^{***} (df = 3; 416)$	

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

#### 4.2 Published Tables: Stargazer2

```
stargazer(fm1,fm3,fm4,title="Regression Results2", add.lines = TRUE)
```

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu % Date and time: 周一, 10 月 22, 2018 - 23:35:23

#### 4.3 Published Tables: Stargazer3

```
stargazer(fm1,fm3,fm4, title="Regression Results3", ci = TRUE)
```

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu % Date and time: 周一, 10 月 22, 2018 - 23:35:24

表 2: Regression Results2

	Dependent variable:			
	testscr			
	(1)	(2)	(3)	
str	$-2.280^{***}$	-1.117***	$-1.129^{***}$	
	(0.480)	(0.240)	(0.247)	
meal_pct		-0.600***	$-0.600^{***}$	
-		(0.017)	(0.017)	
computer			0.0002	
•			(0.001)	
Constant	698.933***	702.911***	703.074***	
	(9.467)	(4.700)	(4.775)	
TRUE				
Observations	420	420	420	
$\mathbb{R}^2$	0.051	0.767	0.767	
Adjusted $\mathbb{R}^2$	0.049	0.766	0.765	
Residual Std. Error	18.581 (df = 418)	9.222 (df = 417)	9.233 (df = 416)	
F Statistic	$22.575^{***} (df = 1; 418)$	$685.756^{***} (df = 2; 417)$	$456.132^{***} (df = 3; 416)$	

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

表 3: Regression Results3

	Dependent variable:  testscr			
	(1)	(2)	(3)	
$\operatorname{str}$	$-2.280^{***}$	-1.117***	-1.129***	
	(-3.220, -1.339)	(-1.588, -0.646)	(-1.613, -0.644)	
${ m meal\_pct}$		-0.600***	-0.600***	
— <b>.</b>		(-0.633, -0.567)	(-0.633, -0.567)	
computer			0.0002	
			(-0.002,0.002)	
Constant	698.933***	702.911***	703.074***	
	(680.377, 717.489)	(693.699, 712.124)	(693.715, 712.433)	
Observations	420	420	420	
$\mathbb{R}^2$	0.051	0.767	0.767	
Adjusted R <sup>2</sup>	0.049	0.766	0.765	
Residual Std. Error	18.581 (df = 418)	9.222 (df = 417)	9.233 (df = 416)	
F Statistic	$22.575^{***} (df = 1; 418)$	$685.756^{***} (df = 2; 417)$	$456.132^{***} (df = 3; 416)$	

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01