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## Overview of R

- Over the past decade, companies have been using a variety of statistical packages, including SPSS, SAS, Minitab, Stata, and others.
- Recently, companies have begun to migrate to a powerful statistical and data mining package called **R**.
- R is an open source software.
  - Free
  - Cost of other packages has driven new users to R.
- R is a language, but we will mostly operate it using a graphical user interface.

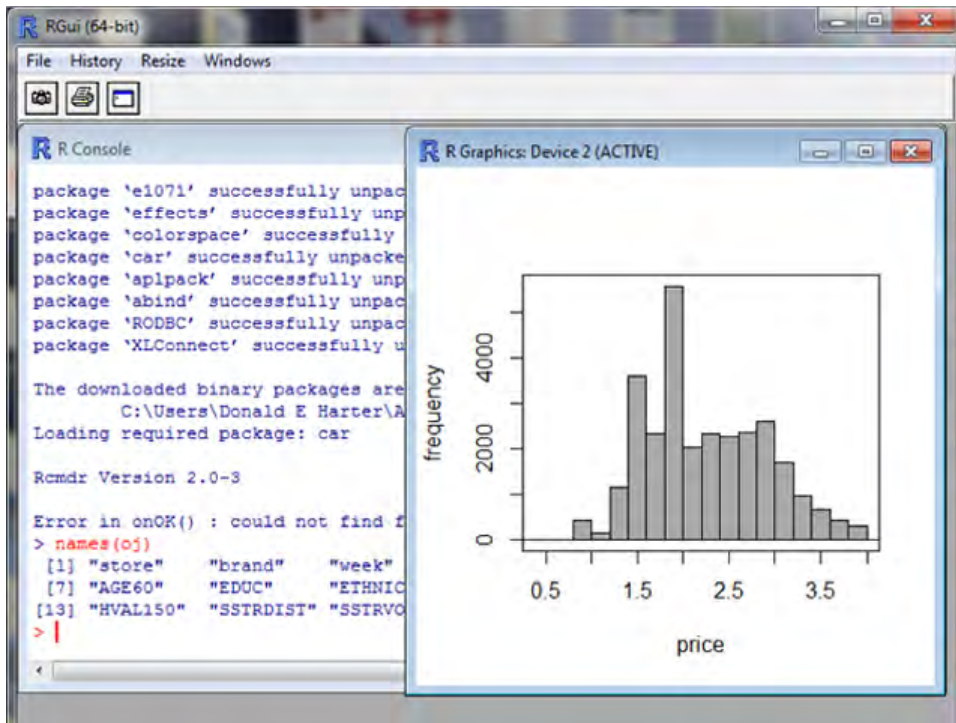
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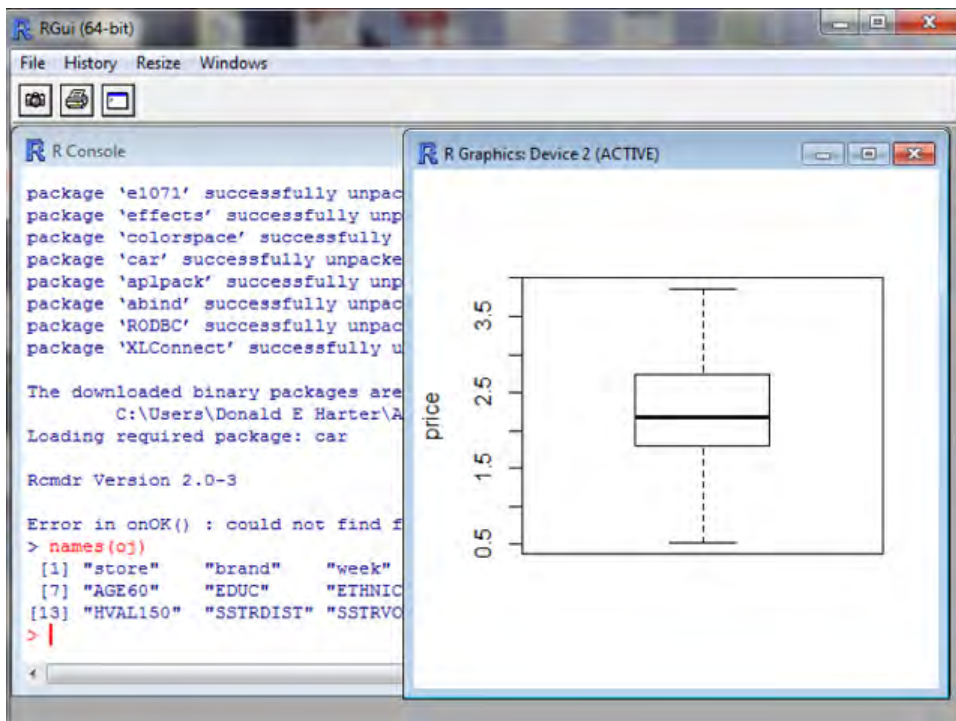
## Visualization

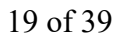
- R has the ability to generate histograms, boxplots, scatterplots, and XY plots similar to Excel.
- Graphs from R can differentiate by brand, product category, or other variable.
  - Gives greater insight into patterns

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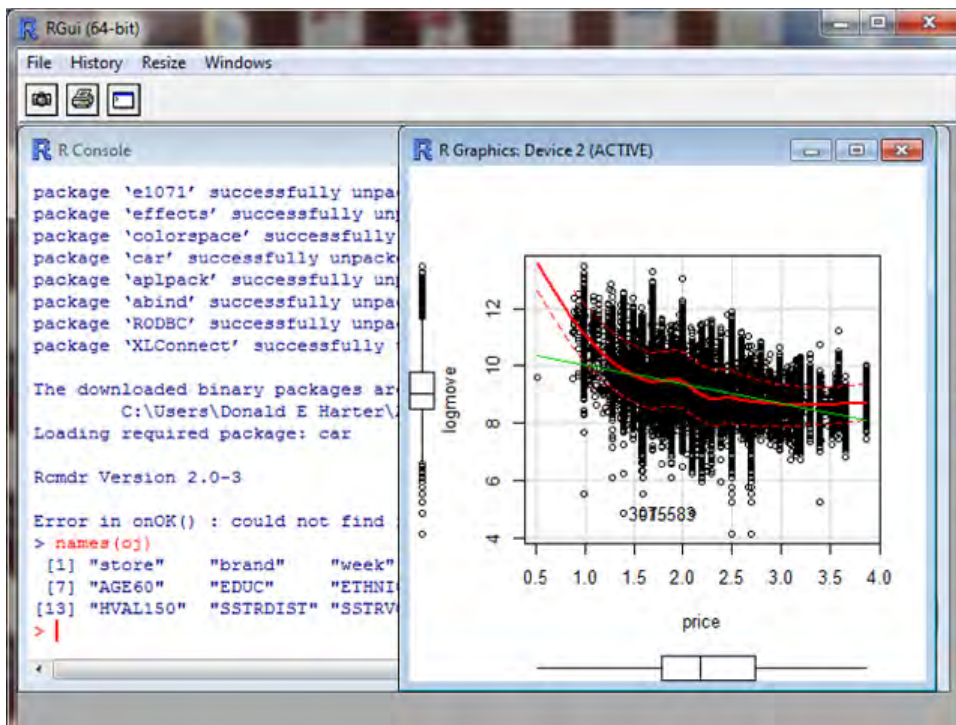
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# Histograms and Boxplots

- **Histograms** show frequency of data within intervals or bins
  - Easy to develop in Excel or R.
- **Boxplots** identify:
  - Maximum and minimum—whiskers
  - 25%-ile and 75%-ile—box
  - Median—center bar
- Boxplots in R can also differentiate by brands.

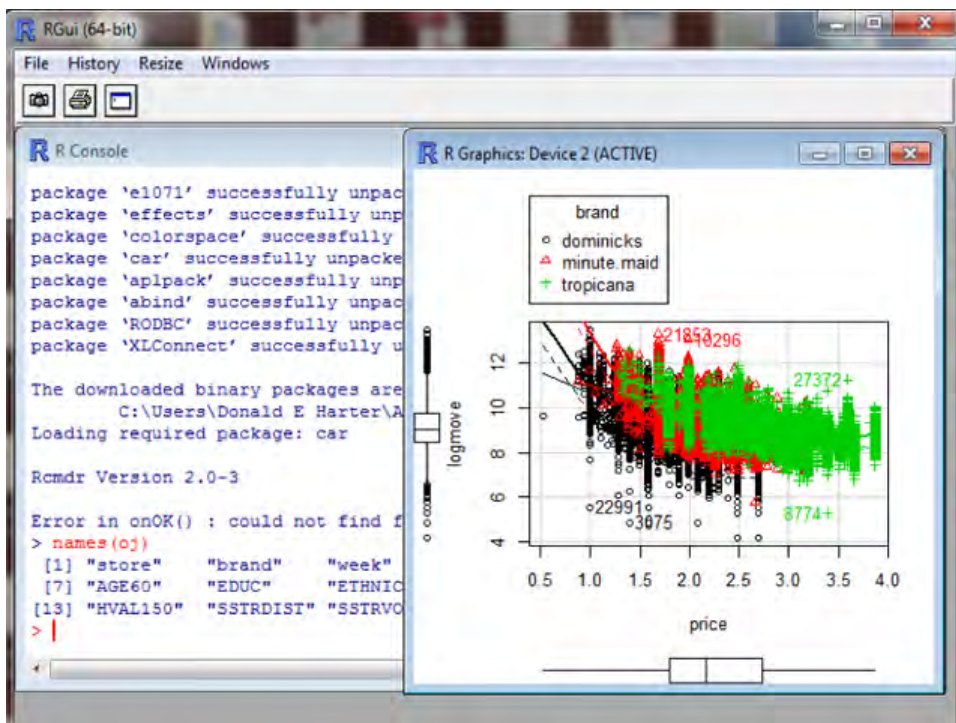


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## Scatterplots

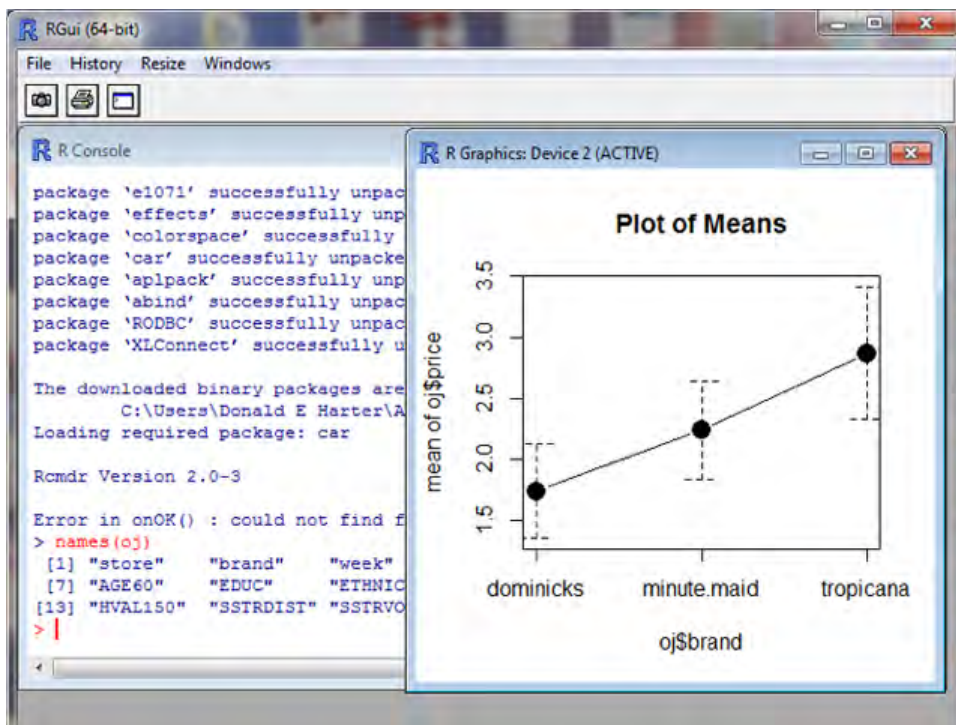
- **Scatterplots** offer not only trend line, but price by price.
- From example:
  - Solid red line—average sales by price
  - Green line—trend line
  - Dotted red line—standard deviation
- Scatterplots of log sales by price can also be categorized by brand.

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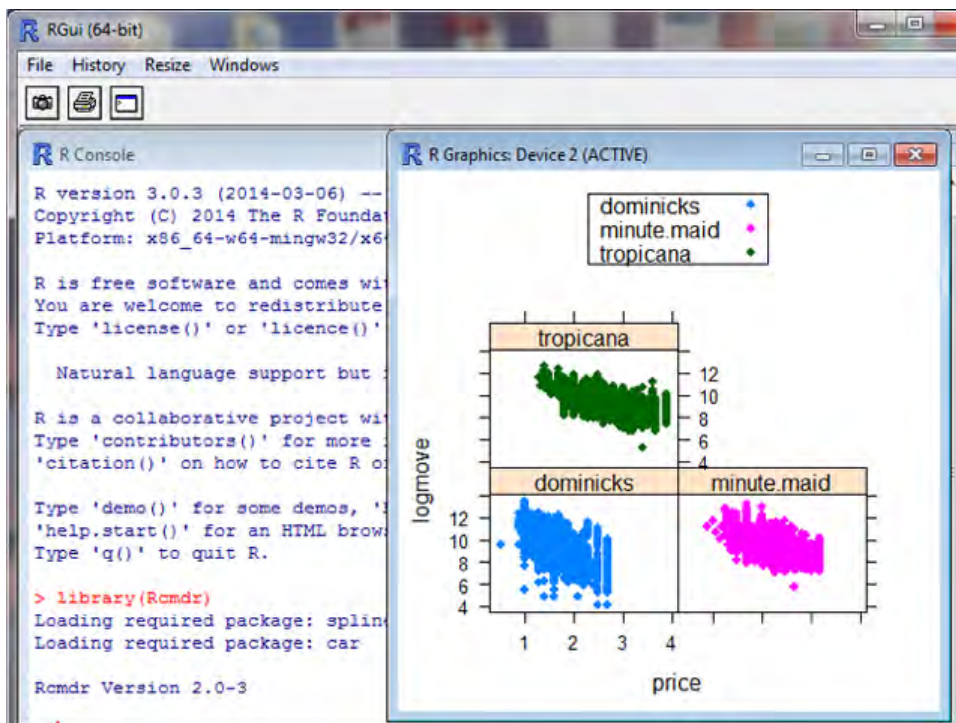




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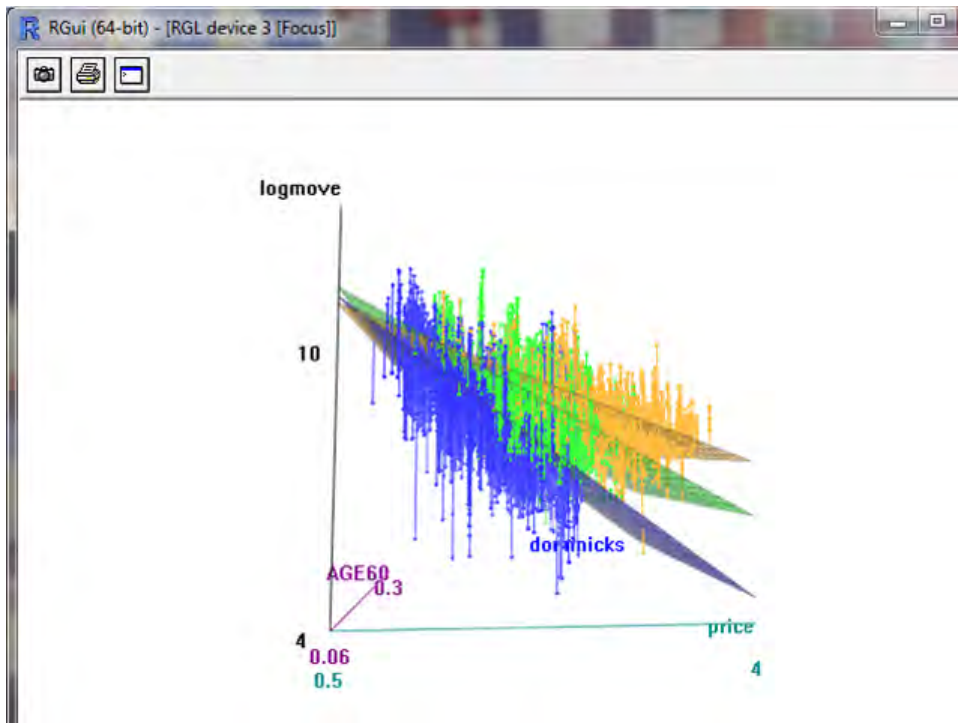
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## Plot of Means and XY Plots

- **Plot of means** is where the average price by brand is plotted.
- **XY plots** display the effect of price on sales per brand by separating and placing plots next to each other.

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### 3-D Visualization

- R can perform more sophisticated 3-D graphing than Excel.
  - Plot the planes—helps find patterns
  - Rotate the graphs—offers different senses of parameters.
- 3-D allows you to see regression lines and planes to identify patterns.

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The screenshot shows the R Commander window with the following content:

**R Script**

```
numSummary(oj[,c("logmove", "price")], statistics=c("mean", "sd", "IQR",
"quantiles"), quantiles=c(0.25,.5,.75,1))
numSummary(oj[,c("logmove", "price")], groups=oj$brand, statistics=c("mean",
"sd", "IQR", "quantiles"), quantiles=c(0.25,.5,.75,1))
numSummary(oj[,c("logmove", "price")], groups=oj$brand, statistics=c("mean",
"sd", "IQR", "quantiles"), quantiles=c(0.25,.5,.75,1))
```

**Output**

Variable: logmove

	mean	sd	IQR	0%	25%	50%	75%
dominicks	9.174831	1.1929370	1.5619512	4.158883	8.392990	9.121728	9.954941
minute.maid	9.217278	0.9852867	1.3523928	5.768321	8.476371	9.026418	9.828764
tropicana	9.111483	0.8473800	0.9685592	5.257495	8.565602	8.987197	9.534161

100% n

	100%	n
dominicks	13.48202	9649
minute.maid	13.29018	9649
tropicana	12.57205	9649

Variable: price

	mean	sd	IQR	0%	25%	50%	75%	100%	n
dominicks	1.735809	0.3858380	0.41	0.52	1.58	1.59	1.99	2.69	9649
minute.maid	2.241162	0.4045146	0.50	0.88	1.99	2.17	2.49	3.17	9649
tropicana	2.870493	0.5485578	0.70	1.29	2.49	2.99	3.19	3.87	9649

**Messages**

```
[3] NOTE: The dataset oj has 28947 rows and 17 columns.
[4] ERROR: You must select a variable.
```

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## Statistical Summaries

- Similar to Excel, R can calculate descriptive statistics.
- R has the additional ability to calculate descriptive statistics by brand.
- Stratify and characterize data while performing analysis.

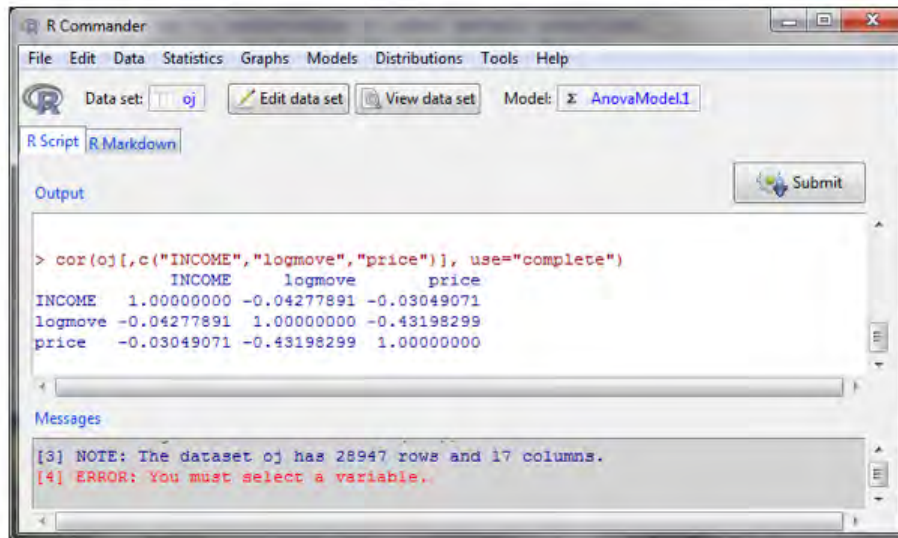
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## Correlations

- Measure how two variables are related
- Positively correlated: one variable increases, the other one also increases
- Negatively correlated: one variable increases, the other one decreases

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The screenshot shows the R Commander window. The 'Data set' dropdown is set to 'oj'. The 'Model' dropdown is set to 'AnovaModel1'. The 'Output' pane displays the following R code and its result:

```
> cor(oj[,c("INCOME", "logmove", "price")], use="complete")
```

	INCOME	logmove	price
INCOME	1.00000000	-0.04277891	-0.03049071
logmove	-0.04277891	1.00000000	-0.43198299
price	-0.03049071	-0.43198299	1.00000000

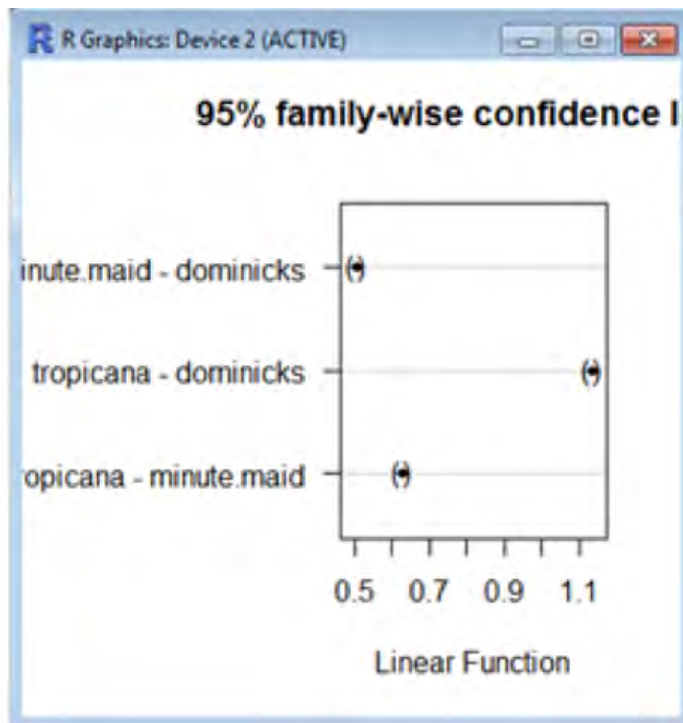
The 'Messages' pane shows the following output:

```
[3] NOTE: The dataset oj has 28947 rows and 17 columns.  
[4] ERROR: You must select a variable.
```

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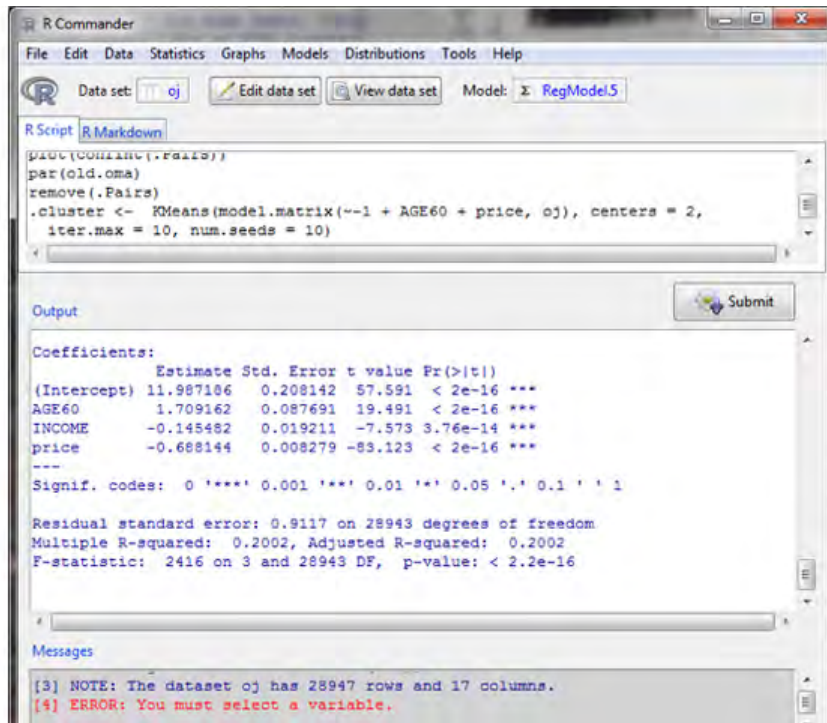
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## Analysis of Variance

- Store managers want to compare product placement positions, sales, and prices of different products or brands.
- Analysis of variance (ANOVA) is used to compare the averages for different categories or brands.
- ANOVA calculates the average of each item, combines with standard deviation/variance, and determines if they are statistically different from one another.
- ANOVA is a valuable technique for product placement.

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The screenshot shows the R Commander window. The 'R Script' pane contains the following code:

```
plot(CONFINT(.Pairs))
par(old.cma)
remove(.Pairs)
.cluster <- KMeans(model.matrix(~-1 + AGE60 + price, oj), centers = 2,
  iter.max = 10, num.seeds = 10)
```

The 'Output' pane displays the results of the regression analysis:

```
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept) 11.987106   0.208142  57.591 < 2e-16 ***
AGE60        1.709162   0.087691  19.491 < 2e-16 ***
INCOME       -0.145482   0.019211  -7.573 3.76e-14 ***
price        -0.688144   0.008279 -83.123 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.9117 on 28943 degrees of freedom
Multiple R-squared:  0.2002, Adjusted R-squared:  0.2002
F-statistic: 2416 on 3 and 28943 DF, p-value: < 2.2e-16
```

The 'Messages' pane shows the following messages:

```
[3] NOTE: The dataset oj has 28947 rows and 17 columns.
[4] ERROR: You must select a variable.
```

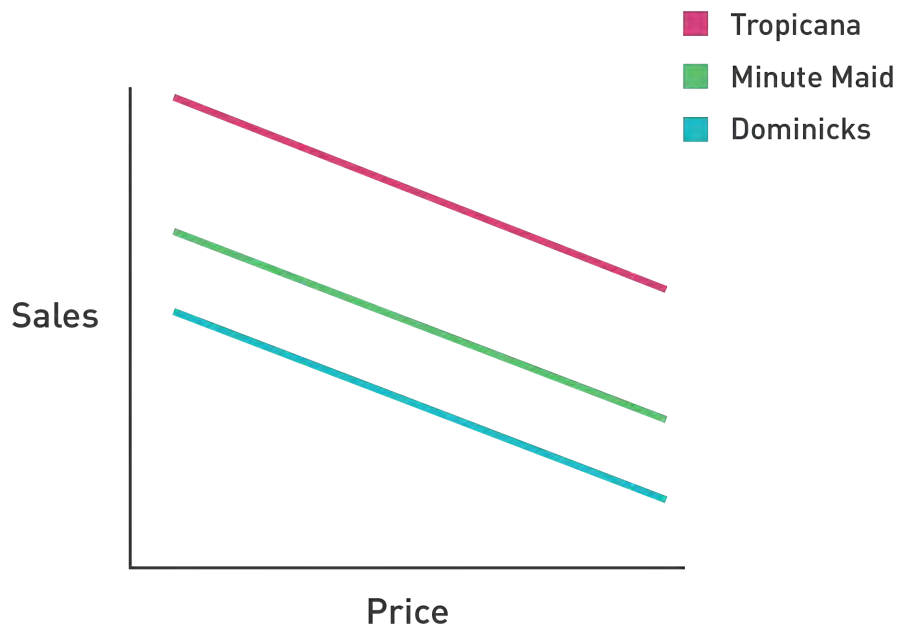
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## Regression

- R can perform linear regression.
  - Output is similar to Excel's multivariate regression output
- Analysis produces coefficients, statistical significance, and R-squared values.
  - Sets foundation to view more complex regressions

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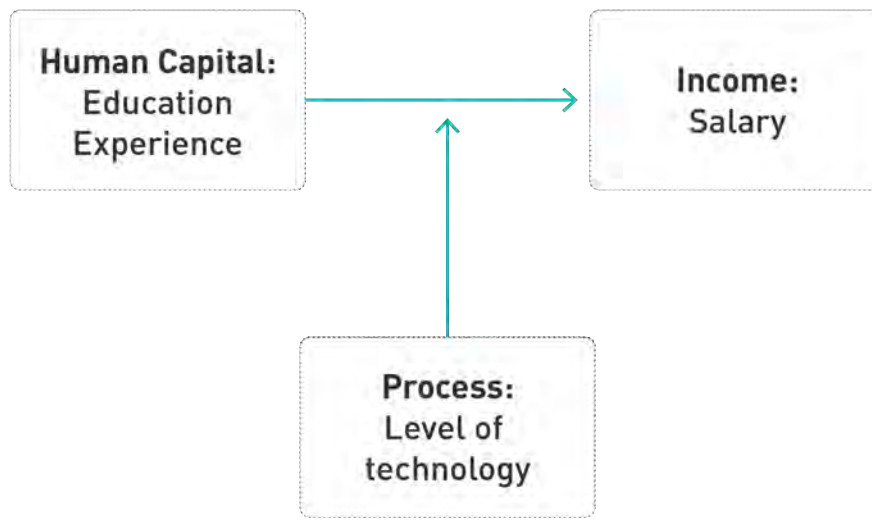
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### Regression with Dummy Variables

- Dummy variables change the intercept of some items to be different from others.
  - E.g., one brand will always be higher in sales than another
  - Reflected in different intercepts or heights of the regression line
- This allows you to look at price differentiation in different product markets.

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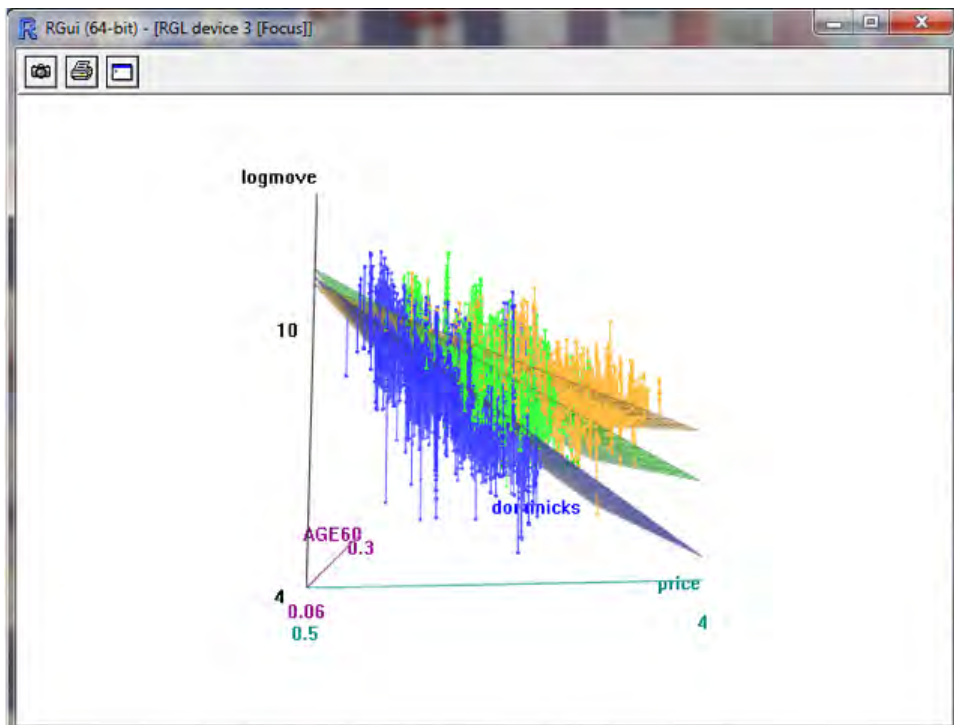


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## Regression with Moderating Effects

- Dummy variables can only change the intercepts.
- Moderating effects allow the slope of the line to change.
  - Recognizes that two variables might interact, magnifying their effects
- Education and experience both have a positive effect on income.
  - When an experienced, educated professional learns technology, the technology is the moderating effect.
- **Moderating effect:** An interaction of variables that leverage each other
- The moderating effect acts as a catalyst to accelerate the effect of certain variables.

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