Why Stata is the best programming language to start data analysis

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What are we comparing?

- Programming language
- Software application
- 3 Documentation
- 4 Community

Programming language

- designed for data analysis
- human readable syntax
- good default options

Two-Column Slide

Brattle

A typical day for Brattle RAs includes:

- · Combining economic theory and industry knowledge to solve real problems
- . Diving into data, using statistical analyses to extract information from messy data
- Constructing models from a blend of theoretical concepts to answer complex questions
- Reviewing literature and industry trends to understand the debate around key developments
- Conducting statistical analysis and working with data using tools such as Stata, R, Excel or Python
- Auditing and contributing to the creation of financial, economic, and operational models



Key responsibilities:

- Interacting extensively with clients to gain insight into their industry
- · Contributing to development of theoretical and empirical approach
- · Utilising literature to support economic arguments
- · Efficiently conducting empirical analysis using Excel and Stata
- Overseeing the day-to-day running of the project
- Drafting reports summarising analysis
- · Delivering an accurate and high-quality work product
- · Participating actively in client meetings and conference calls
- · Extensive mentoring and supervising of junior staff

```
/* Hotel price data */
use "hotels-europe price.dta", clear
/* Add hotel features (location,
  stars, ratings, etc.) */
merge m:1 hotel_id using
  "hotels-europe features.dta"
/* Censor prices that are too high */
replace price = 1000 if price > 1000
/* Regress price on ratings, stars.
  plus month, weekend dummies */
regress price rating stars i.month
  i.weekend, vce(cluster country)
```

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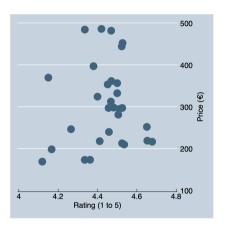
Linear regression	Number of obs	=	115,367
	F(10, 30)	=	272.88
	Prob > F	=	0.0000
	R-squared	=	0.2577
	Root MSE	-	146 52

(Std. Err. adjusted for 31 clusters in country)

Robust							
price	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval	
rating	21.5814	7.861631	2.75	0.010	5.52581	37.6369	
stars	52.54748	8.304822	6.33	0.000	35.58677	69.508	
month							
2	6.944091	5.554252	1.25	0.221	-4.399204	18.2873	
3	22.07722	5.573216	3.96	0.000	10.6952	33.4592	
4	29.2734	4.929571	5.94	0.000	19.20587	39.3409	
5	40.27256	4.755351	8.47	0.000	30.56084	49.984	
6	40.54402	5.855406	6.92	0.000	28.58568	52.5023	
11	9.108877	4.401348	2.07	0.047	.1201249	18.0976	
12	187.1044	15.04021	12.44	0.000	156.3882	217.82	
1.weekend	1.828793	6.036309	0.30	0.764	-10.49899	14.156	
_cons	-142.8199	16.73315	-8.54	0.000	-176.9935	-108.646	

```
/* keep only 5-star hotels */
keep if stars == 5
/ * mean price and rating by country */
collapse (mean) price (mean) rating,
  by(country)
label variable price "Price (€)"
label variable rating "Rating (1 to 5)"
scatter price rating, scheme(economist)
```

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Stata vs R

Stata vs Python

```
replace price = 1000 if price > 1000 data.loc[data["price"] > 1000, "price"]
```

Same in Python

```
import pandas as pd
import matplotlib.pvplot as plt
# load hotel price data
price data = pd.read stata("hotels-europe price.dta")
# add hotel features (location, stars, ratings, etc.)
features = pd.read stata("hotels-europe features.dta")
data = price data.merge(features, on="hotel id", how="left")
# replace high prices with 1000
data.loc[data["price"] > 1000, "price"] = 1000
# regress price on ratings, stars, plus month, weekend dummies
data = pd.get dummies(data, columns=["month", "weekend"])
result = sm.OLS(data["price"], data[["rating", "stars"] + list(data.columns[data.columns.str.startswith("month_")])
 + list(data.columns[data.columns.str.startswith("weekend_")])]).fit(cov_type="cluster", cov_kwds={"groups": data["country"]})
# keep only 5-star hotels
data = data[data["stars"] == 5]
# calculate mean price and rating by country
data = data.groupby("country").mean()[["price", "rating"]]
# label variables
data.rename(columns={"price": "Price (€)", "rating": "Rating (1 to 5)"}, inplace=True)
# scatterplot
data.plot(x="Price (€)", v="Rating (1 to 5)", kind="scatter", colormap="tab10", figsize=(8, 6))
plt.show()
```

Same in R

```
library(tidyverse)
library(ggplot2)
# load hotel price data
price data <- read dta("hotels-europe price.dta")
# add hotel features (location, stars, ratings, etc.)
features <- read dta("hotels-europe features.dta")</pre>
data <- left join(price data, features, by="hotel id")
# replace high prices with 1000
data <- data %>% mutate(price=if else(price > 1000, 1000, price))
# regress price on ratings, stars, plus month, weekend dummies
data <- data %>% mutate(month=factor(month), weekend=factor(weekend)) %>% nest(-country)
result <- data %>% mutate(model=map(data, ~ lm(price ~ rating + stars + month + weekend, data=.)),
                         summ=map(model, broom::tidy)) %>%
                unnest(summ)
# subset data for 5-star hotels only
five star data <- data %>% filter(stars == 5) %>%
                        group by(country) %>%
                        summarize(mean price=mean(price), mean rating=mean(rating))
# create scatterplot
ggplot(five_star_data, aes(x=mean_price, y=mean_rating)) +
  geom point() +
  labs(x="Price (€)", v="Rating (1 to 5)") +
  scale color economist()
```