

What is Quantile Regression and How to use the Rpackage quantreg

Quantile Regression (Part 2)

Lecture Video Slides

Engel Dataset from Rpackage quantreg

- Dataset that records Family Expenditure on Food and Family Income in Belgium 1857.
- Used to show a limitation of Linear Regression and usefulness of Quantile Regression.
- Dataset is in quantreg Rpackage.

	income	foodexp
1	420.1577	255.8394
2	541.4117	310.9587
3	901.1575	485.6800
4	639.0802	402.9974
5	750.8756	495.5608
6	945.7989	633.7978
7	829.3979	630.7566
8	979.1648	700.4409
9	1309.8789	830.9586
10	1492.3987	815.3602

First 10 of 235 records in Engel Dataset.

R Package quantreg

- Install R Package quantreg
- `library(quantreg)`
- The main function for quantile regression is `rq()`
- `?rq` at Rconsole to view the documentation.

```
> ?rq  
> |
```

Quantile Regression

Description

Returns an object of class "rq" "rqs" or "rq.process" that represents a quantile regression fit.

Usage

```
rq(formula, tau=.5, data, subset, weights, na.action,  
   method="br", model = TRUE, contrasts, ...)
```

Arguments

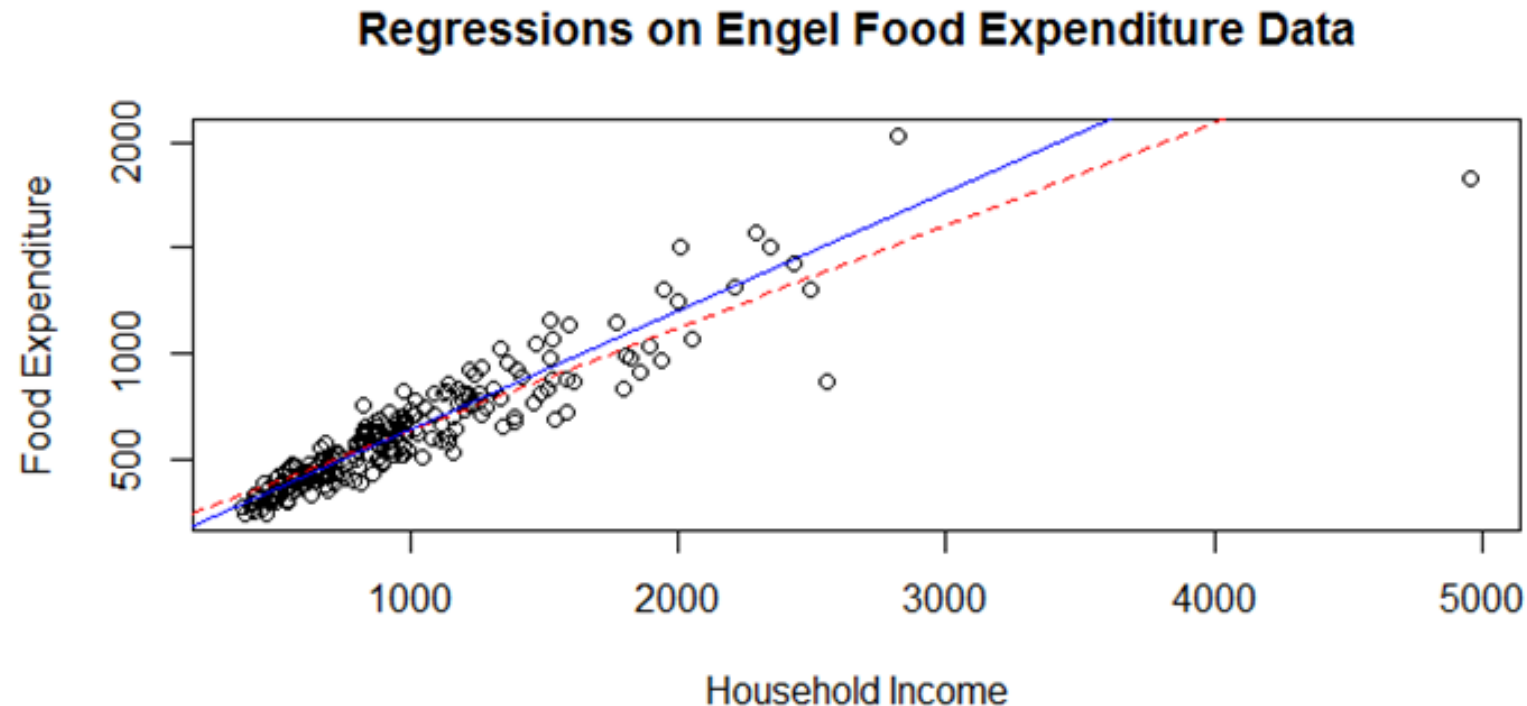
formula a formula object, with the response on the left of a ~ operator, and the terms, separated by + operators, on the right.

tau the quantile(s) to be estimated, this is generally a number strictly between 0 and 1, but if

Parameter tau in rq() function controls the Percentile of Y

```
library(quantreg)
# Fit 50th Percentile Line (i.e. Median)
fit.p.5 <- rq(engel$foodexp ~ engel$income, tau=.5)
abline(fit.p.5, col="blue")
```

Parameter tau = 0.5



Median represented as Blue Line.
Mean represented as dotted red line.

Run the name of the model to get model coefficients

```
> fit.p.5  
Call:  
rq(formula = foodexp ~ income, tau = 0.5)  
  
Coefficients:  
(Intercept)      income  
  81.4822474    0.5601806  
  
Degrees of freedom: 235 total; 233 residual
```

50th Percentile: Food Expenditure = 81.48 + 0.56*income

Use summary() to get model coefficients and confidence intervals

```
> summary(fit.p.5)

Call: rq(formula = foodexp ~ income, tau = 0.5)

tau: [1] 0.5

Coefficients:
              coefficients lower bd  upper bd
(Intercept)  81.48225      53.25915 114.01156
income        0.56018      0.48702   0.60199
```

Q: What is the critical number that we want to check in the 95% confidence interval?

A: Check that the interval excludes 0.

Use `summary(..., se = "nid")` to get p-values

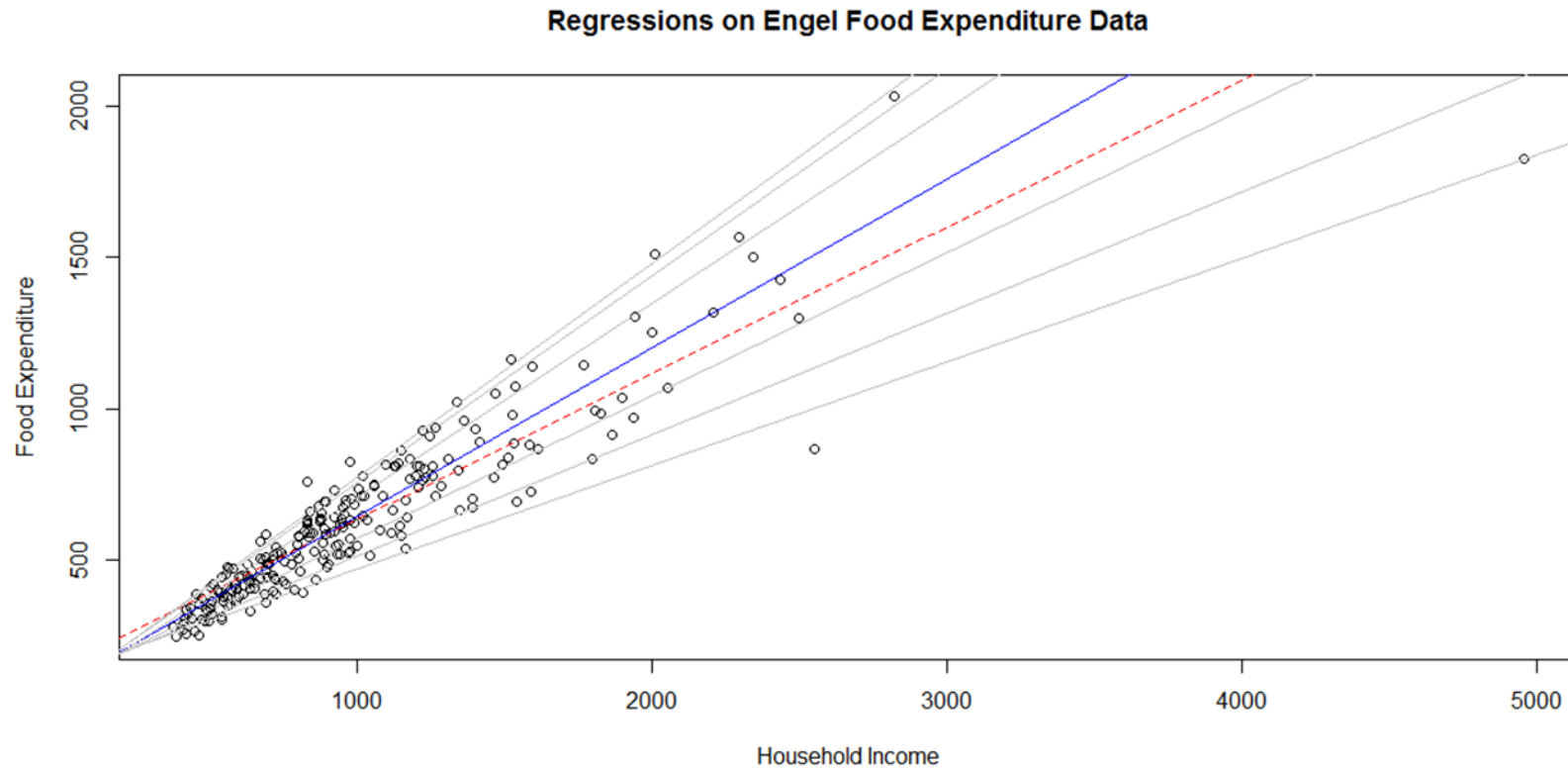
```
> summary(fit.p.5, se = "nid")  
  
call: rq(formula = foodexp ~ income, tau = 0.5)  
  
tau: [1] 0.5  
  
Coefficients:  
              Value      Std. Error t value  Pr(>|t|)  
(Intercept) 81.48225    19.25066     4.23270 0.00003  
income       0.56018     0.02828    19.81032 0.00000
```

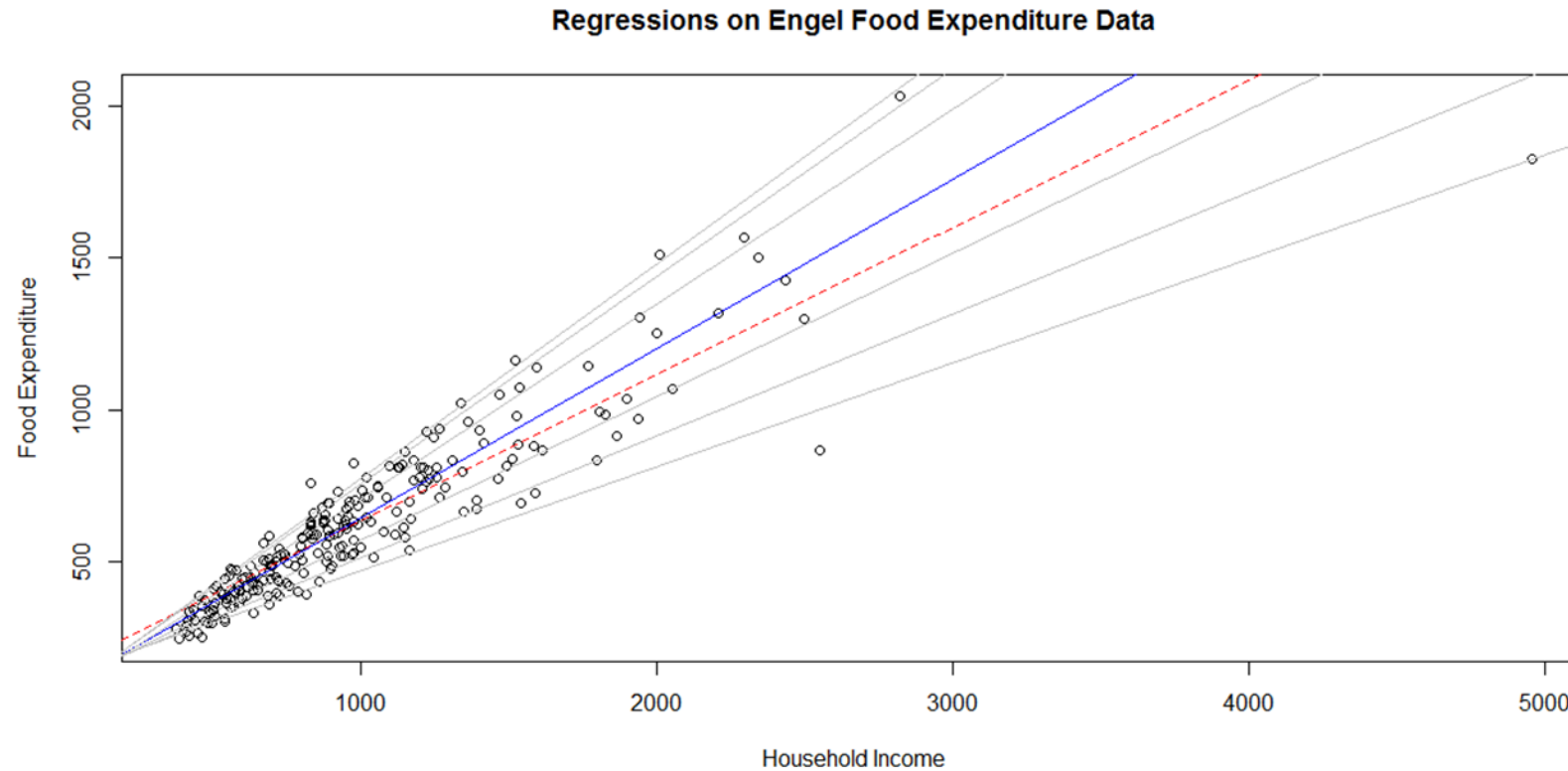
Q: What is the cut-off that we want to check in the p-values?

A: Check p-value < 0.05.

Fit & Plot Six Percentiles with a for loop

```
# 5th, 10th, 25th, 75th, 90th, 95th percentiles.  
taus <- c(.05, .1, .25, .75, .90, .95)  
  
# Plot the 6 percentile grey lines  
for( i in 1:length(taus)){  
  abline(rq(engel$foodexp~engel$income,tau=taus[i]), col = "grey")  
}
```





Q: How does the quantile regression lines contribute to the data analysis?

A: They quantify the relationship between income and food expenditure at various percentiles. They show precisely how the variance in food expenditure increases as income increases.

Quantile Regression in Python

- statsmodels
 - Includes Engel dataset
 - https://www.statsmodels.org/dev/examples/notebooks/generated/quantile_regression.html
- scikit-learn
 - https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.QuantileRegressor.html

sklearn.linear_model.QuantileRegressor

```
class sklearn.linear_model.QuantileRegressor(*, quantile=0.5, alpha=1.0, fit_intercept=True, solver='interior-point', solver_options=None)
```

How were Quantreg “lines” determined?

- Check Function.
- Hands-on Exercise.