



Quantile Regression

Presented by
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Quantile Regression

- Constructs linear models for predicting specified quantiles.
- Useful when:
 - primary interest concerns a percentile of the distribution rather than the mean.
 - the distribution of the data at a specified combination of the predictor variables is not Gaussian.
 - the variance of Y depends on X .
 - there are outliers present.

Applications

- Growth curves
- Ecology
- Epidemiology
- Health services utilization
- CEO pay
- Household income
- Home prices
- Sea ice extent
- Astrophysics
- Chemistry
- Genomics
- Waiting times
- Product reliability

Basic Model Structure

$$Q_{\tau}(Y) = \beta_0(\tau) + \beta_1(\tau)X_1 + \beta_2(\tau)X_2 + \cdots + \beta_p(\tau)X_p + \epsilon$$

$Q_{\tau}(Y)$: conditional τ -th quantile of dependent variable Y

X_1, X_2, \dots, X_p : predictor variables

Note that the coefficients depend on τ .

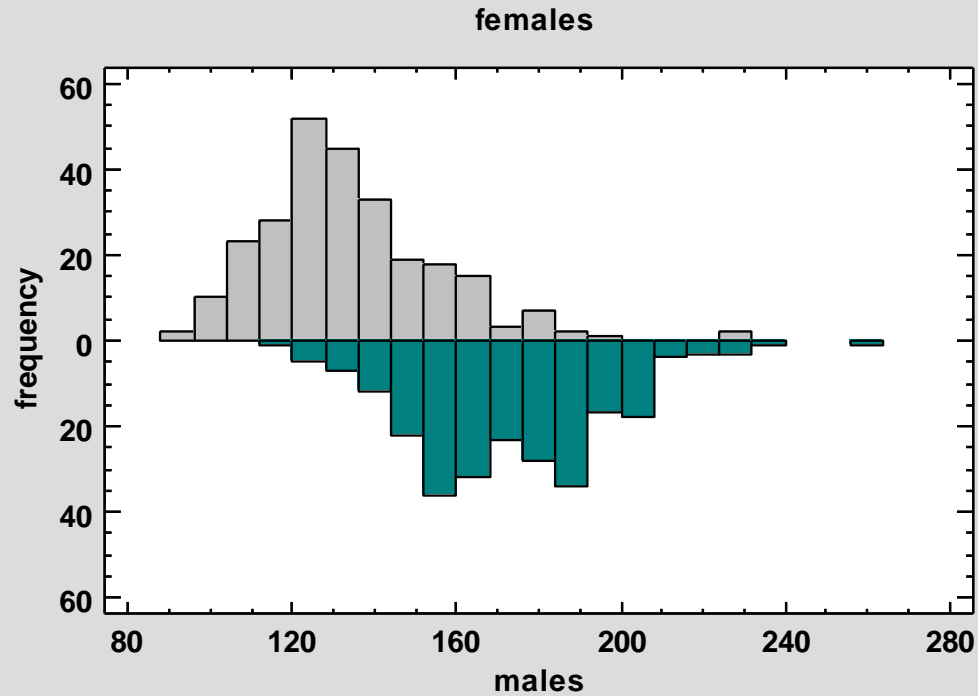
Statgraphics

- Statgraphics uses the *quantreg* program in R to fit models.
- *Quantreg* was written by R. Koenker.
- You should download the latest build (19.2.02) which includes a few tweaks to the *Quantile Regression* procedure.

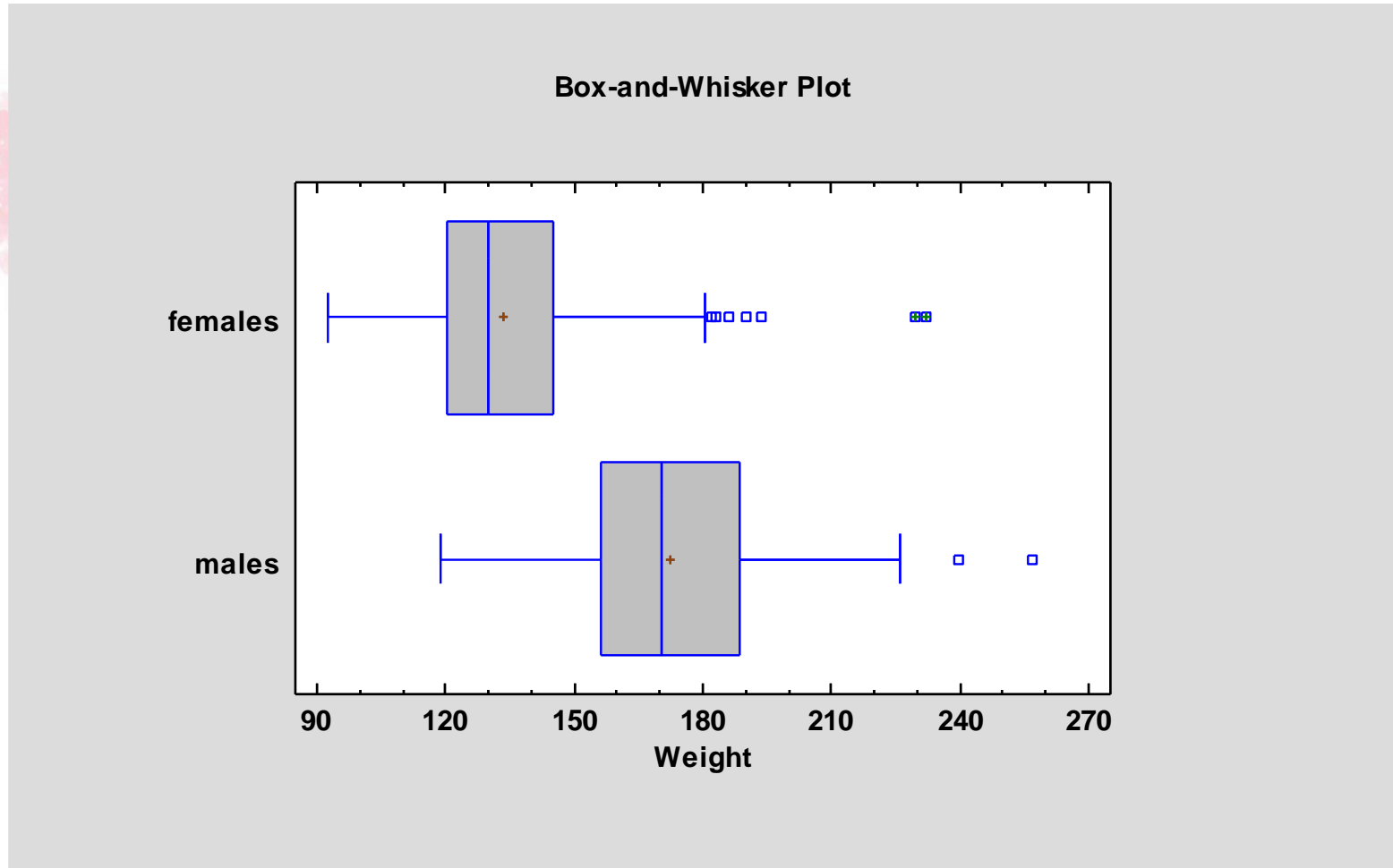
Example #1

- First example is taken from the Journal of Statistics Education Data Archive.
- Information about 247 men and 267 women sampled at fitness centers in California:
 - 21 body dimension measurements
 - age
 - height
 - weight
 - gender

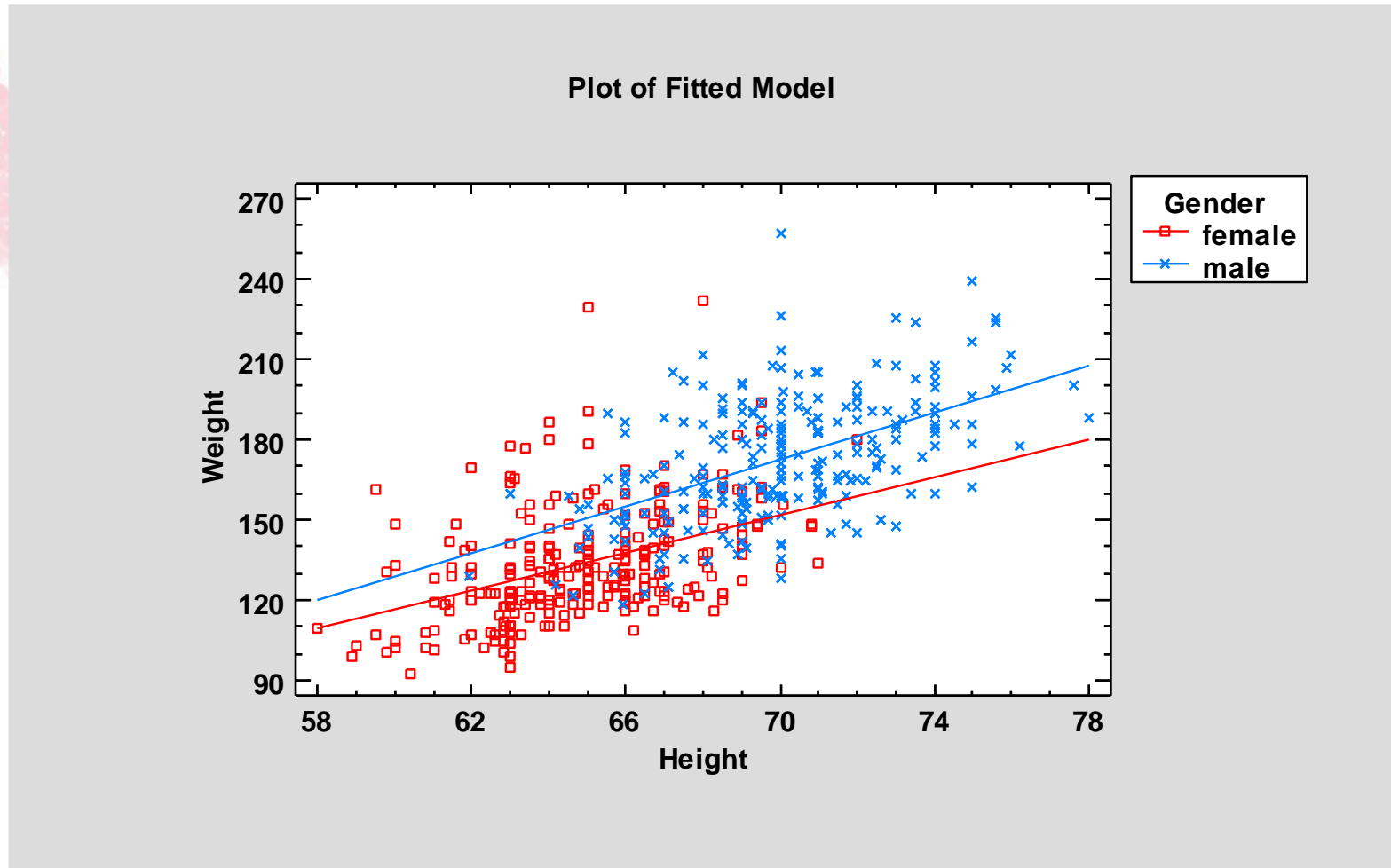
Histogram of Weight



Box and Whisker Plot of Weight



Comparison of Regression Lines



Data Input Dialog Box

Quantile Regression

Dependent Variable:
▶ Weight

Categorical Factors:
▶ Gender

Quantitative Factors:
▶ Height
▶ Age

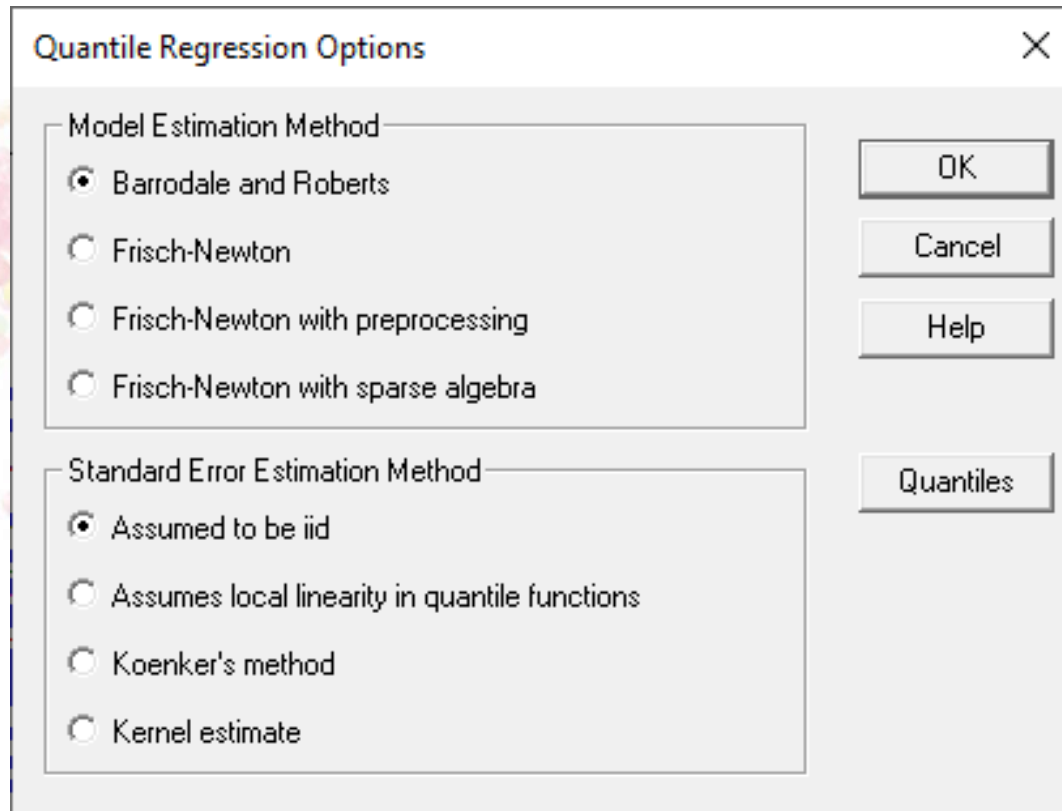
(Weights:)
▶

(Select:)
▶

Sort column names

OK Cancel Delete Transform... Help

Analysis Options Dialog Box



The image shows a dialog box titled "Quantile Regression Options" with a close button (X) in the top right corner. The dialog is divided into two main sections, each with a title and a list of radio button options. On the right side of the dialog, there are four buttons: "OK", "Cancel", "Help", and "Quantiles".

Quantile Regression Options

Model Estimation Method

- Barrodale and Roberts
- Frisch-Newton
- Frisch-Newton with preprocessing
- Frisch-Newton with sparse algebra

Standard Error Estimation Method

- Assumed to be iid
- Assumes local linearity in quantile functions
- Koenker's method
- Kernel estimate

Buttons: OK, Cancel, Help, Quantiles

Methods

- *Barrodale and Roberts* – for up to several thousand observations.
- *Frisch-Newton* – useful for larger problems.
- *Frisch-Newton with preprocessing* – useful for even larger problems. Best for large n and small p .
- *Frisch-Newton with sparse algebra* – useful for large n and large p .

Quantile Dialog Box

Quantile Regression Quantile Levels ✕

0.05			OK Cancel Help
0.1			
0.25			
0.5			
0.75			
0.9			
0.95			

Tables and Graphs

Tables and Graphs

TABLES	GRAPHS
<input checked="" type="checkbox"/> Analysis Summary	<input checked="" type="checkbox"/> Quantile Plot
<input checked="" type="checkbox"/> Estimated Quantiles	<input checked="" type="checkbox"/> Coefficient Plot
<input type="checkbox"/> Predicted Quantiles	<input type="checkbox"/> Residual Scatterplot
<input type="checkbox"/> Residuals	<input type="checkbox"/> Residual Box-and-Whisker Plot
<input type="checkbox"/> R Script and Messages	<input type="checkbox"/> Residual Density Trace

OK
Cancel
All
Store
Help

Analysis Summary

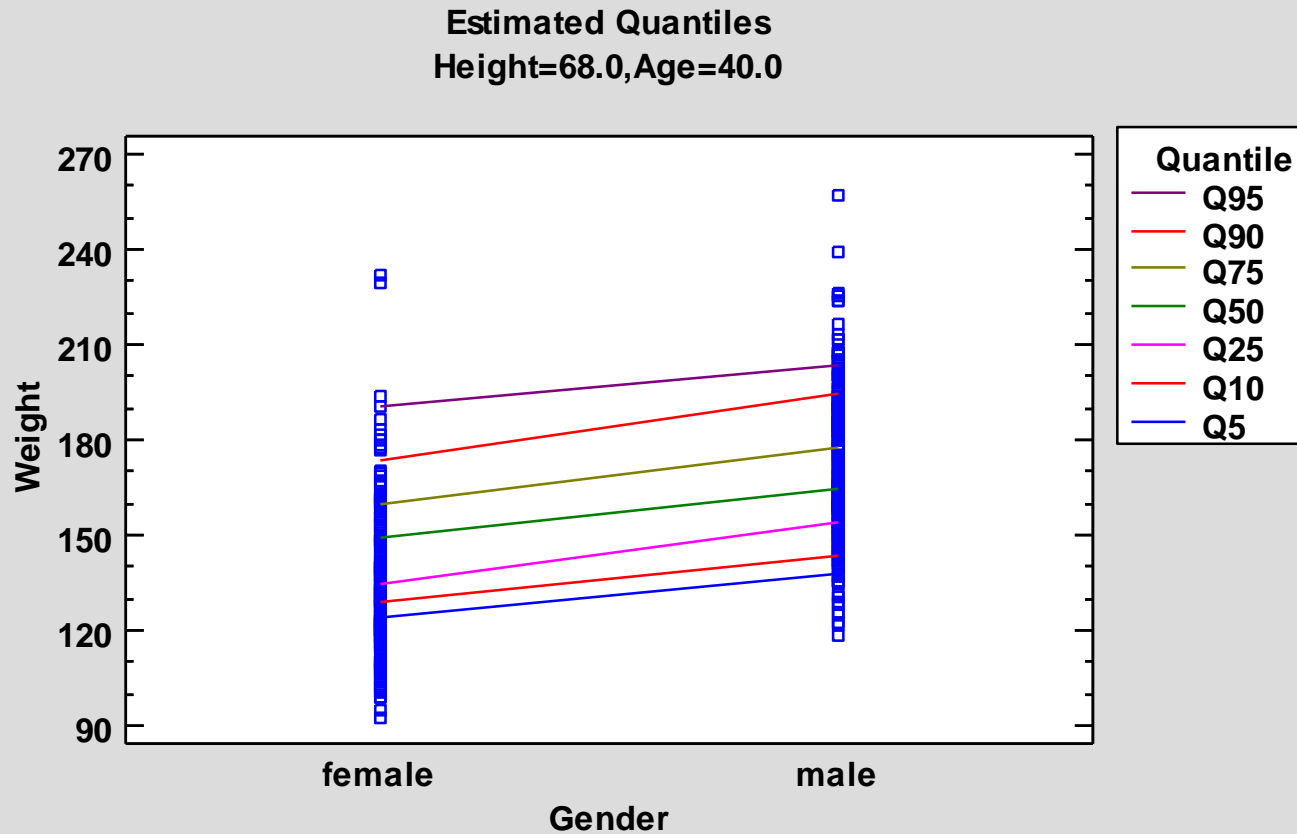
```
Quantile Regression - Weight

##
## Call: rq(formula = Weight ~ as.factor(Gender) + Height + Age, tau = c(0.05,
##      0.1, 0.25, 0.5, 0.75, 0.9, 0.95), data = d, na.action = na.omit,
##      method = "br")
##
## tau: [1] 0.05
##
## Coefficients:
##              Value      Std. Error t value  Pr(>|t|)
## (Intercept) -126.82664    34.12032   -3.71704  0.00022
## as.factor(Gender)1    13.61688     3.86453    3.52355  0.00046
## Height          3.54535     0.51748    6.85119  0.00000
## Age              0.24501     0.14704    1.66626  0.09628
##
```

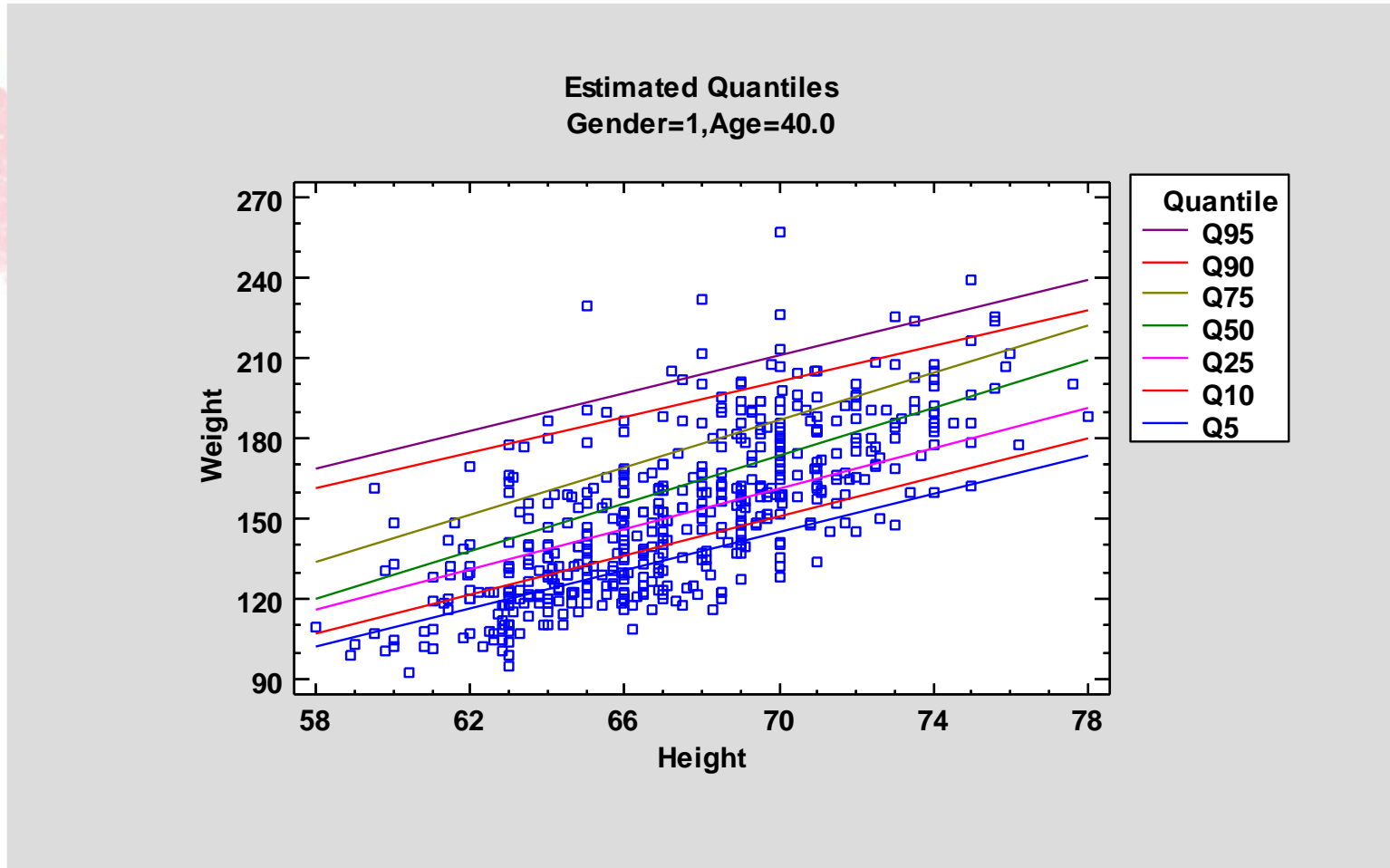

Analysis Summary

```
Quantile Regression - Weight
##
## Call: rq(formula = Weight ~ as.factor(Gender) + Height + Age, tau = c(0.05,
##      0.1, 0.25, 0.5, 0.75, 0.9, 0.95), data = d, na.action = na.omit,
##      method = "br")
##
## tau: [1] 0.5
##
## Coefficients:
##              Value      Std. Error t value    Pr(>|t|)
## (Intercept)  -169.75731    23.59772   -7.19380  0.00000
## as.factor(Gender)1  15.73834     2.67272    5.88851  0.00000
## Height         4.45059     0.35789   12.43564  0.00000
## Age            0.39921     0.10169    3.92564  0.00010
##
```

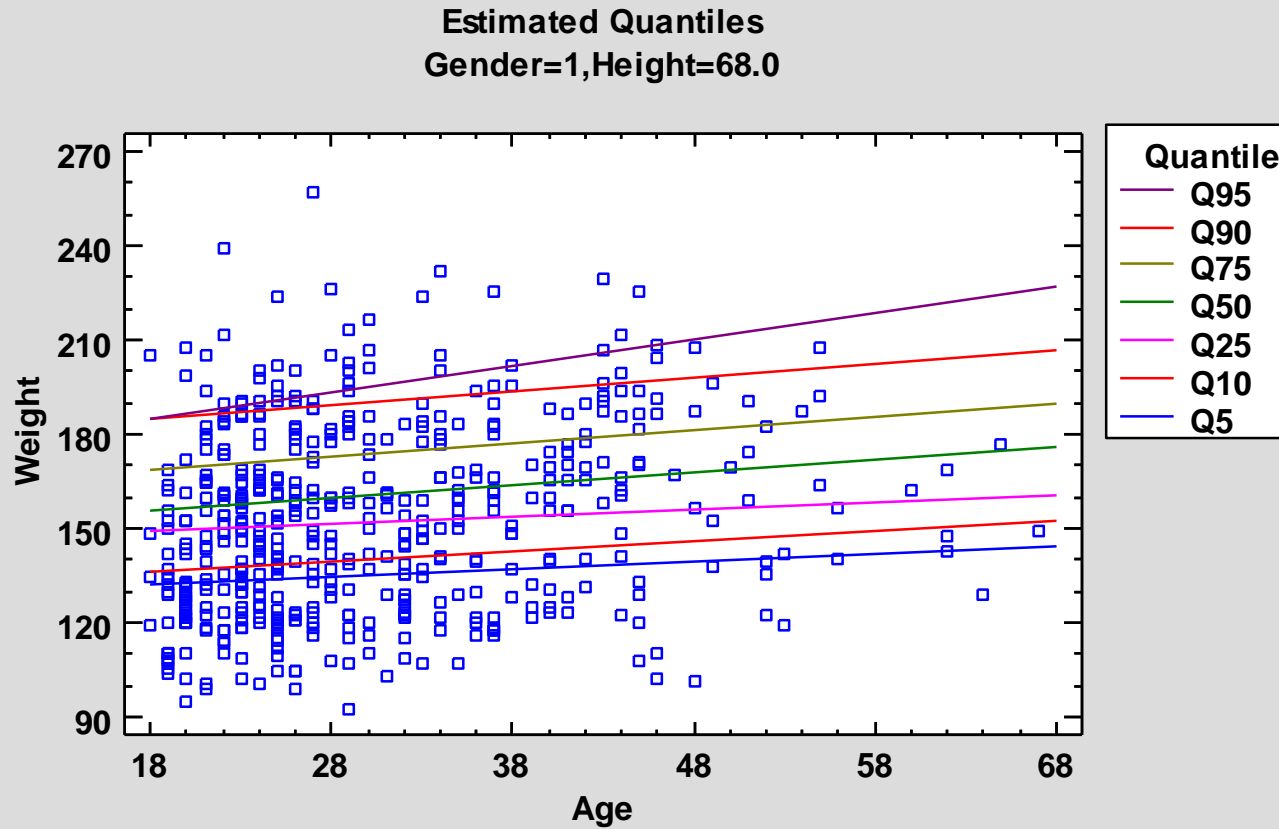
Effect of Gender



Effect of Height



Effect of Age



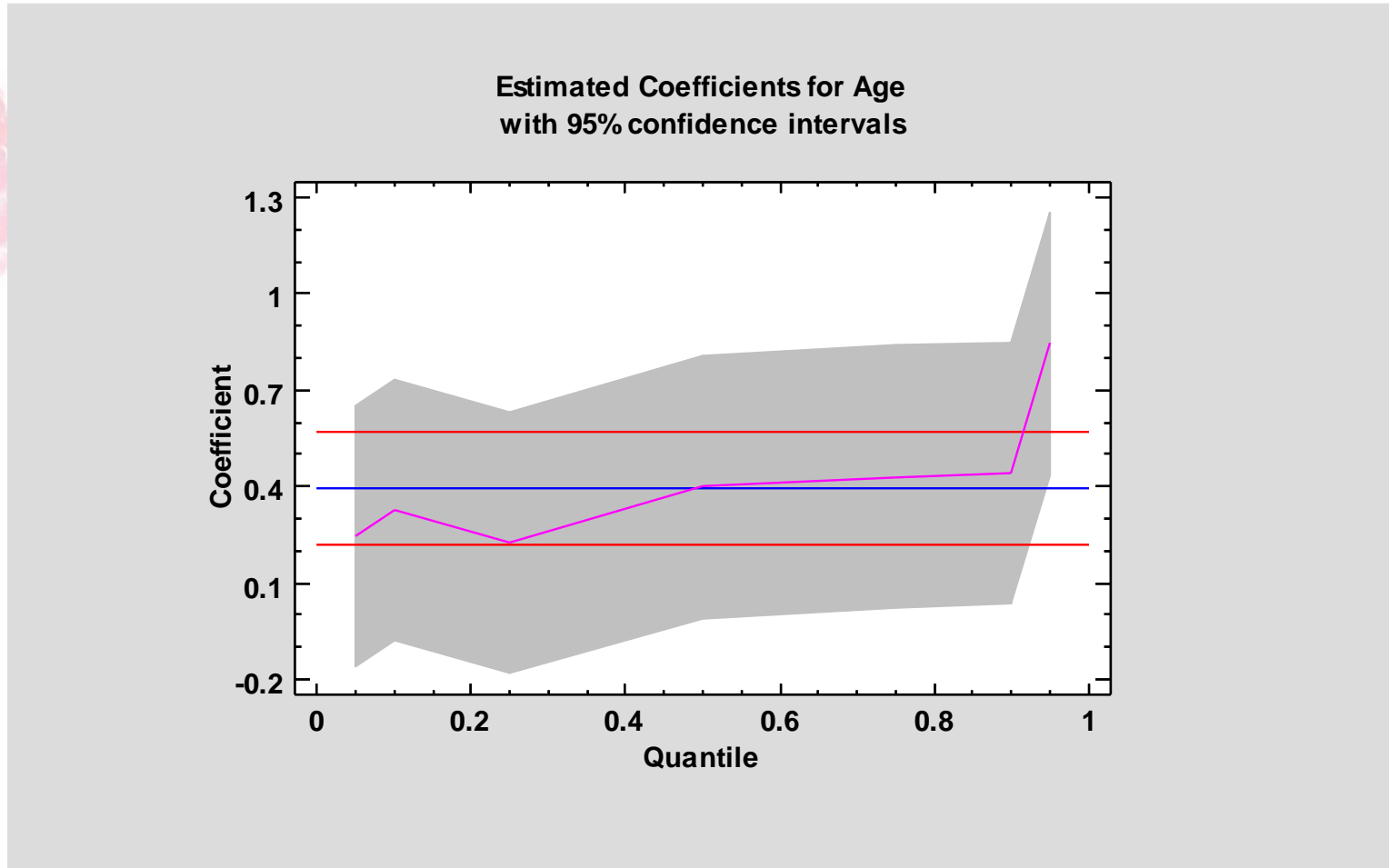
Estimated Quantiles

Quantile Regression - Weight

Estimated Quantiles

Row	Weight	Gender	Height	Age	Q5	Q10	Q25	Q50	Q75	Q90	Q95
1	144.6	1.0	68.5	21.0	134.792	139.228	151.483	159.23	171.951	187.948	189.403
2	158.3	1.0	69.0	23.0	137.055	141.721	153.809	162.254	175.023	190.5	192.862
3	177.9	1.0	76.2	28.0	163.806	169.811	181.966	196.294	209.008	216.707	222.57
4	160.1	1.0	73.4	23.0	152.654	157.884	170.33	181.836	194.475	205.163	208.436
5	173.8	1.0	73.7	22.0	153.473	158.658	171.232	182.772	195.371	205.72	208.654
6	164.9	1.0	71.5	21.0	145.428	150.248	162.747	172.582	185.214	197.946	200.022
7	190.5	1.0	72.4	26.0	149.844	155.196	167.249	178.583	191.347	203.159	207.43
8	172.9	1.0	72.6	27.0	150.798	156.258	168.225	179.873	192.662	204.268	208.983
9	136.7	1.0	68.9	23.0	136.7	141.354	153.434	161.809	174.581	190.167	192.508
10	179.9	1.0	72.4	21.0	148.619	153.554	166.126	176.587	189.192	200.945	203.208
11	168.9	1.0	70.9	23.0	143.791	148.701	160.943	170.71	183.423	196.832	199.587
12	184.3	1.0	70.0	22.0	140.355	145.067	157.34	166.305	179.013	193.39	195.557
13	198.5	1.0	75.6	20.0	159.719	164.981	177.917	190.43	202.909	211.167	213.69
14	164.5	1.0	69.3	26.0	138.853	143.808	155.609	164.787	177.642	192.828	196.457
15	156.6	1.0	68.5	23.0	135.282	139.885	151.932	160.028	172.813	188.834	191.092
16	175.5	1.0	72.4	22.0	148.864	153.883	166.351	176.987	189.623	201.388	204.052
17	206.8	1.0	75.9	30.0	163.232	169.365	181.289	195.757	208.544	216.593	223.197
18	154.4	1.0	67.5	22.0	131.492	135.883	147.953	155.179	167.961	185.059	186.708
19	159.6	1.0	68.1	29.0	135.334	140.385	151.777	160.643	173.63	190.156	194.743
20	189.4	1.0	69.3	22.0	137.873	142.495	154.711	163.19	175.918	191.057	193.079
21	173.8	1.0	69.3	22.0	137.873	142.495	154.711	163.19	175.918	191.057	193.079

Comparing Quantile Coefficients

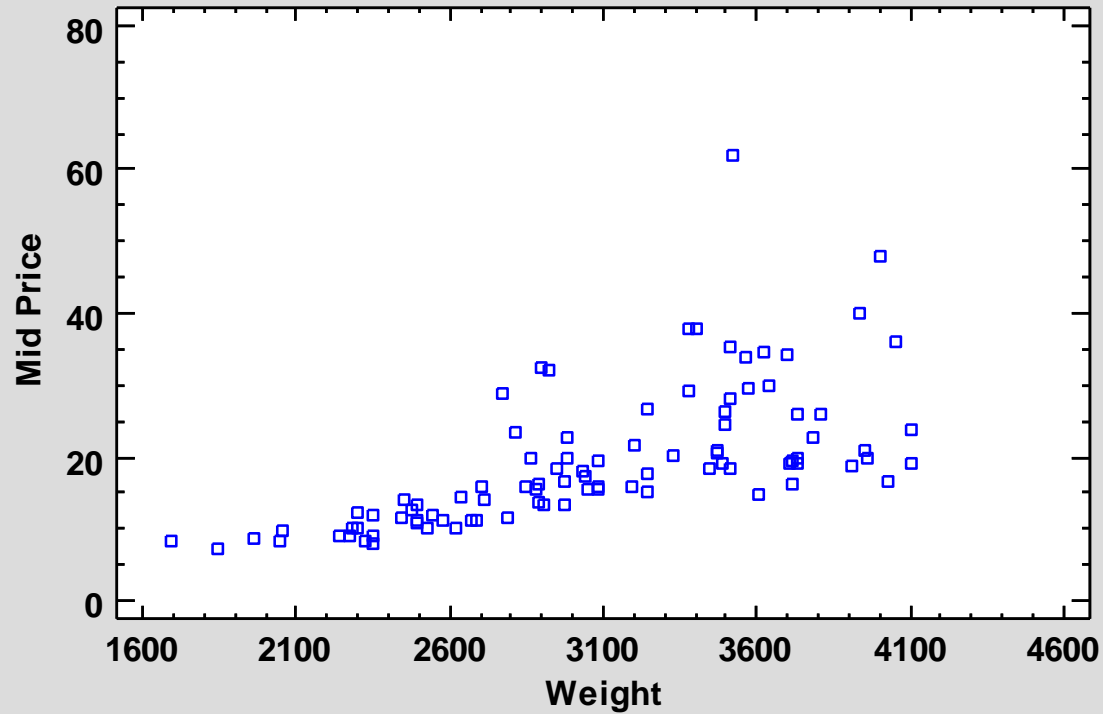


Example #2

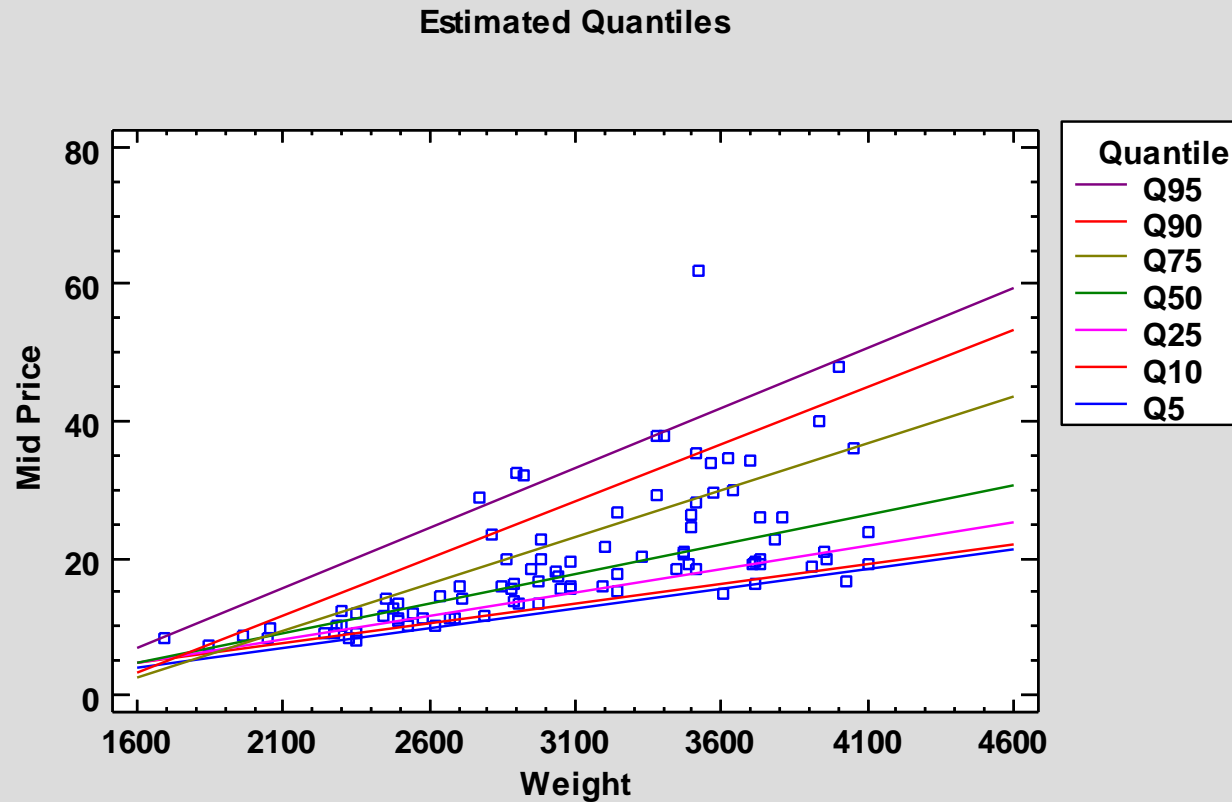
- Second example is also taken from the Journal of Statistics Education Data Archive.
- Information about 93 makes and models of automobiles manufactured in 1993
 - Price of automobile
 - Weight

Scatterplot

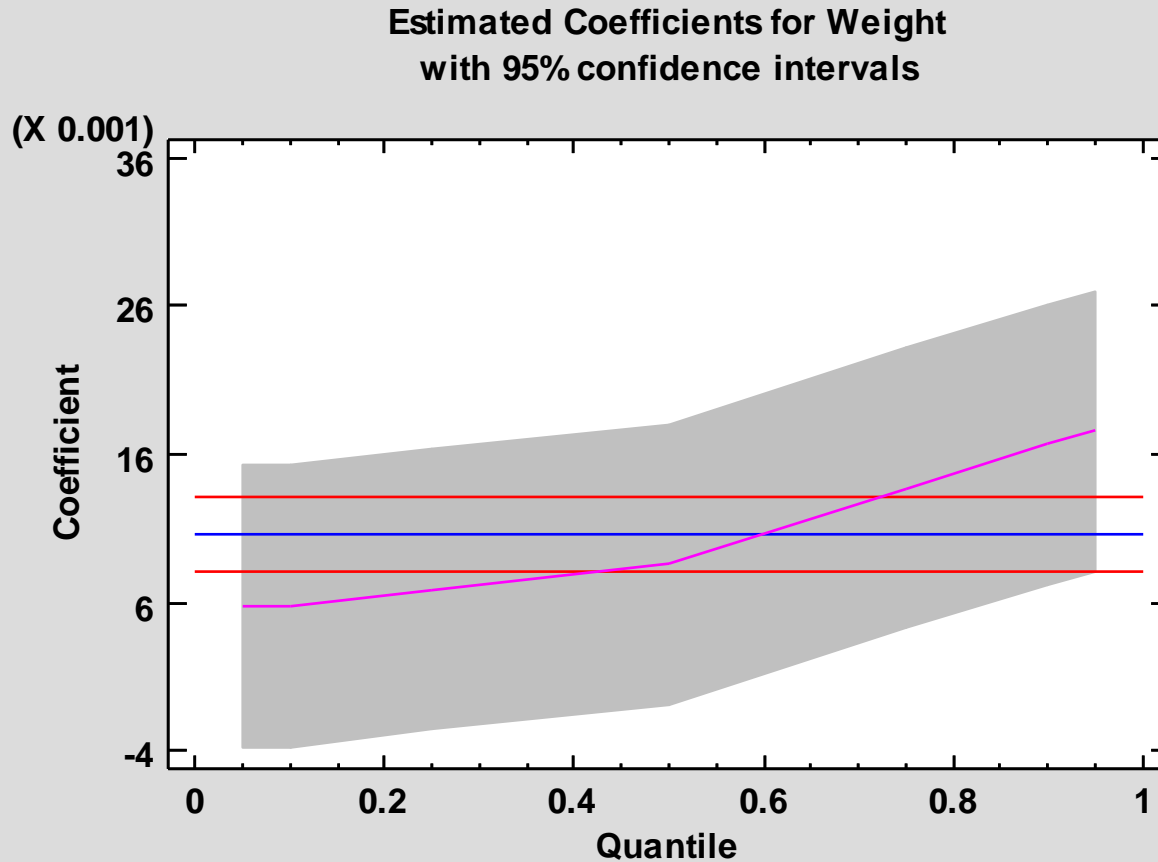
Plot of Mid Price vs Weight



Estimated Quantiles



Comparing Quantile Coefficients



References

- StatFolios and data files are at:
www.statgraphics.com/webinars
- R Package “quantreg” (2021) <https://cran.r-project.org/web/packages/quantreg/quantreg.pdf>
- Quantile Regression (2005) Roger Koenker.
(Econometric Society Monographs, Series Number 38)
- Body and car data from:
http://jse.amstat.org/jse_data_archive.htm