

The short answer is that you interpret quantile regression coefficients just like you do ordinary regression coefficients. The long answer is that you interpret quantile regression coefficients almost just like ordinary regression coefficients.

We can illustrate this with a couple of examples using the **hsb2** dataset.

```
use https://stats.idre.ucla.edu/stat/stata/notes/hsb2, clear
```

```
tabstat write, by(female) stat(p25 p50 p75)
```

```
Summary for variables: write
by categories of: female
```

| female | p25 | p50 | p75 |
|--------|------|-----|-----|
| male | 41 | 52 | 59 |
| female | 50 | 57 | 62 |
| Total | 45.5 | 54 | 60 |

We will begin by running median and .75 quantile regression models without any predictors.

```
qreg write
```

```
Iteration 1: WLS sum of weighted deviations = 1595.95
```

| | | | | | | | |
|--|--|-------------------|-----------|-------|-----------------|----------------------|----------|
| Iteration 1: sum of abs. weighted deviations = | | | | | 1591 | | |
| Iteration 2: sum of abs. weighted deviations = | | | | | 1571 | | |
| Median regression | | | | | Number of obs = | | 200 |
| Raw sum of deviations | | 1571 (about 54) | | | | | |
| Min sum of deviations | | 1571 | | | Pseudo R2 = | | 0.0000 |
| | | | | | | | |
| write | | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] | |
| | | | | | | | |
| _cons | | 54 | 1.239519 | 43.57 | 0.000 | 51.55572 | 56.44428 |
| | | | | | | | |
| qreg write, quantile(.75) | | | | | | | |
| Iteration 1: WLS sum of weighted deviations = | | | | | 1237.9502 | | |
| Iteration 1: sum of abs. weighted deviations = | | | | | 1202.5 | | |
| Iteration 2: sum of abs. weighted deviations = | | | | | 1084.5 | | |
| 75 Quantile regression | | | | | Number of obs = | | 200 |
| Raw sum of deviations | | 1084.5 (about 60) | | | | | |
| Min sum of deviations | | 1084.5 | | | Pseudo R2 = | | 0.0000 |
| | | | | | | | |
| write | | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] | |
| | | | | | | | |
| _cons | | 60 | .6665574 | 90.01 | 0.000 | 58.68558 | 61.31442 |
| | | | | | | | |

In the median regression the constant is the median of the sample while in the .75 quantile regression the constant is the 75th percentile for the sample.

Next, we'll add the binary predictor **female** to the model.

```
qreg write female
```

```
Iteration 1: WLS sum of weighted deviations = 1543.9433
```

```
Iteration 1: sum of abs. weighted deviations = 1545
Iteration 2: sum of abs. weighted deviations = 1542
Iteration 3: sum of abs. weighted deviations = 1536
```

```
Median regression
Raw sum of deviations 1571 (about 54)
Min sum of deviations 1536
Number of obs = 200
Pseudo R2 = 0.0223
```

| write | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] |
|--------|-------|-----------|-------|-------|----------------------|
| female | 5 | 2.611711 | 1.91 | 0.057 | -.1503394 10.15034 |
| _cons | 52 | 1.927268 | 26.98 | 0.000 | 48.19939 55.80061 |

```
predict p50
(option xb assumed; fitted values)
```

```
tabulate p50
```

| Fitted values | Freq. | Percent | Cum. |
|---------------|-------|---------|--------|
| 52 | 91 | 45.50 | 45.50 |
| 57 | 109 | 54.50 | 100.00 |
| Total | 200 | 100.00 | |

```
qreg write female, quantile(.75)
```

```
Iteration 1: WLS sum of weighted deviations = 1204.3893
```

```
Iteration 1: sum of abs. weighted deviations = 1272
Iteration 2: sum of abs. weighted deviations = 1154.5
Iteration 3: sum of abs. weighted deviations = 1060
```

```
75 Quantile regression
Raw sum of deviations 1084.5 (about 60)
Min sum of deviations 1060
Number of obs = 200
Pseudo R2 = 0.0226
```

| write | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] |
|--------|-------|-----------|------|-------|----------------------|
| female | 3 | 1.23163 | 2.44 | 0.016 | .5712035 5.428796 |

| | | | | | | |
|------------------------------------|-----------|-----------------|--------------|--------------|-----------------|-----------------|
| _cons | 59 | .9385943 | 62.86 | 0.000 | 57.14908 | 60.85092 |
| <hr/> | | | | | | |
| predict p75 | | | | | | |
| (option xb assumed; fitted values) | | | | | | |
| tabulate p75 | | | | | | |
| Fitted values | | Freq. | Percent | | Cum. | |
| <hr/> | | | | | | |
| 59 | | 91 | 45.50 | | 45.50 | |
| 62 | | 109 | 54.50 | | 100.00 | |
| <hr/> | | | | | | |
| Total | | 200 | 100.00 | | | |
| <hr/> | | | | | | |

From this point on I'll describe what is going on in the median regression model. The interpretation for the .75 quantile regression is basically the same except that you substitute the term 75th percentile for the term median.

With the binary predictor, the constant is median for group coded zero (males) and the coefficient is the difference in medians between males and female (see the **tabstat** above).

Looking at the tabulated predicted scores we see that we get two values, the conditional median for males (52) and the conditional median for female (57).

Now, let me show you something that is really neat about quantile regression. I will replace the highest value of **write** (67) with the value of 670 and rerun these analyses.

```
replace write=670 if write==67
(7 real changes made)
```

qreg write female

Iteration 1: WLS sum of weighted deviations = 8319.5083

Iteration 1: sum of abs. weighted deviations = 6544

Iteration 2: sum of abs. weighted deviations = 6156

Iteration 3: sum of abs. weighted deviations = 5757

Median regression

Number of obs = 200

Raw sum of deviations 5792 (about 54)

Min sum of deviations 5757

Pseudo R2 = 0.0060

| write | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] |
|--------|-------|-----------|-------|-------|----------------------|
| female | 5 | 2.611711 | 1.91 | 0.057 | -.1503394 10.15034 |
| _cons | 52 | 1.927268 | 26.98 | 0.000 | 48.19939 55.80061 |

qreg write female, quantile(.75)

Iteration 1: WLS sum of weighted deviations = 11445.07

Iteration 1: sum of abs. weighted deviations = 7582

Iteration 2: sum of abs. weighted deviations = 7461

Iteration 3: sum of abs. weighted deviations = 7391.5

75 Quantile regression

Number of obs = 200

Raw sum of deviations 7416 (about 60)

Min sum of deviations 7391.5

Pseudo R2 = 0.0033

| write | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] |
|--------|-------|-----------|-------|-------|----------------------|
| female | 3 | 1.23163 | 2.44 | 0.016 | .5712035 5.428796 |
| _cons | 59 | .9385943 | 62.86 | 0.000 | 57.14908 60.85092 |

Notice that neither the coefficients nor the standard errors changed. This is because changing this extreme score does not change either the median or the 75th percentile. The only changes that affect the results are when a value crosses a quantile boundary. For example, changing a value of 58 to 580 would not affect the median but would affect the 75th percentile.

For the last example, we will reload the data and use a continuous predictor in the model.

```
use https://stats.idre.ucla.edu/stat/stata/notes/hsb2, clear  
qreg write socst
```

Iteration 1: WLS sum of weighted deviations = 1219.9071

Iteration 1: sum of abs. weighted deviations = 1219.9333

Iteration 2: sum of abs. weighted deviations = 1212.8

Iteration 3: sum of abs. weighted deviations = 1212.5667

Iteration 4: sum of abs. weighted deviations = 1209.375

Iteration 5: sum of abs. weighted deviations = 1208.9

Median regression

Number of obs = 200

Raw sum of deviations 1571 (about 54)

Min sum of deviations 1208.9

Pseudo R2 = 0.2305

| write | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] | |
|-------|----------|-----------|-------|-------|----------------------|----------|
| socst | .6333333 | .0571053 | 11.09 | 0.000 | .5207206 | .7459461 |
| _cons | 20.03333 | 3.069487 | 6.53 | 0.000 | 13.98025 | 26.08642 |

predict double p50

(option xb assumed; fitted values)

qreg write socst, quantile(.75)

Iteration 1: WLS sum of weighted deviations = 992.87

Iteration 1: sum of abs. weighted deviations = 1003.2667

Iteration 2: sum of abs. weighted deviations = 950.85

Iteration 3: sum of abs. weighted deviations = 936.30001

Iteration 4: sum of abs. weighted deviations = 928.66667

Iteration 5: sum of abs. weighted deviations = 926.07501

Iteration 6: sum of abs. weighted deviations = 924.30001

75 Quantile regression

Number of obs = 200

Raw sum of deviations 1084.5 (about 60)

Min sum of deviations 924.3

Pseudo R2 = 0.1477

| write | Coef. | Std. Err. | t | P> t | [95% Conf. Interval] | |
|-------|-------|-----------|-------|-------|----------------------|----------|
| socst | .4 | .0408158 | 9.80 | 0.000 | .3195104 | .4804896 |
| _cons | 37.6 | 2.187081 | 17.19 | 0.000 | 33.28704 | 41.91296 |

```
predict double p75
(option xb assumed; fitted values)
```

With the continuous predictor **socst** the constant is the predicted value when **socst** is zero. The quantile regression coefficient tells us that for every one unit change in **socst** that the predicted value of **write** will increase by .633333.

We can show this by listing the predictor with the associated predicted values for two adjacent values. Notice that for the one unit change from 41 to 42 in **socst** the predicted value increases by .633333.

```
sort socst

list socst p50 p75 in 42/43
```

| | socst | p50 | p75 |
|-----|-------|-----------|------|
| 42. | 41 | 46 | 54 |
| 43. | 42 | 46.633333 | 54.4 |

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