









64 CHAPTER Z SIMPLE REGRESSION ANALYSIS

* A POSITIVE R VALUE INDICATES A POSITIVE RELATIONSHIP, MEANING AS X INCREASES, SO DOES Y, A NEGATIVE R VALUE MEANS AS THE X VALUE INCREASES, THE Y VALUE DECREASES.





66 CHAPTER 2 SIMPLE REGRESSION ANALYSIS







STEP 1: DRAW A SCATTER PLOT OF THE INDEPENDENT VARIABLE VERSUS THE DEPENDENT VARIABLE. IF THE DOTS LINE UP, THE VARIABLES MAY BE CORRELATED.



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STEP 2: CALCULATE THE REGRESSION EQUATION.







Stepl

Find

- The sum of squares of x, S_{xx} : $(x \bar{x})^2$
- The sum of squares of y, S_{yy} : $(y \bar{y})^2$
- The sum of products of x and y, S_{xy} : $(x \bar{x})(y \bar{y})$

Note: The bar over a variable (like $\overline{\mathbf{y}}$) is a notation that means average. We can call this variable x-bar.

	High temp.	Iced tea					
	in °C x	orders <i>y</i>	$x - \overline{x}$	$oldsymbol{y} - oldsymbol{ar{y}}$	$(\boldsymbol{x}-\bar{\boldsymbol{x}})^2$	$(\boldsymbol{y}-ar{\boldsymbol{y}})^2$	$(\boldsymbol{x}-\bar{\boldsymbol{x}})(\boldsymbol{y}-\bar{\boldsymbol{y}})$
22nd (Mon.)	29	77	-0.1	4.4	0.0	19.6	-0.6
23rd (Tues.)	28	62	-1.1	-10.6	1.3	111.8	12.1
24th (Wed.)	34	93	4.9	20.4	23.6	417.3	99.2
25th (Thurs.)	31	84	1.9	11.4	3.4	130.6	21.2
26th (Fri.)	25	59	-4.1	-13.6	17.2	184.2	56.2
27th (Sat.)	29	64	-0.1	-8.6	0.0	73.5	1.2
28th (Sun.)	32	80	2.9	7.4	8.2	55.2	21.2
29th (Mon.)	31	75	1.9	2.4	3.4	5.9	4.5
30th (Tues.)	24	58	-5.1	-14.6	26.4	212.3	74.9
31st (Wed.)	33	91	3.9	18.4	14.9	339.6	71.1
1st (Thurs.)	25	51	-4.1	-21.6	17.2	465.3	89.4
2nd (Fri.)	31	73	1.9	0.4	3.4	0.2	0.8
3rd (Sat.)	26	65	-3.1	-7.6	9.9	57.8	23.8
4th (Sun.)	30	84	0.9	11.4	0.7	130.6	9.8
Sum	408	1016	0	0	129.7	2203.4	484.9
Average	29.1	72.6					
		↓	_		¥	↓ ↓	¥
	r	ū			\mathbf{S}_{xx}	\mathbf{S}_{yy}	\mathbf{S}_{xy}

72 CHAPTER 2 SIMPLE REGRESSION ANALYSIS

* SOME OF THE FIGURES IN THIS CHAPTER ARE ROUNDED FOR THE SAKE OF PRINTING, BUT CALCULATIONS ARE DONE USING THE FULL, UNROUNDED VALUES RESULTING FROM THE RAW DATA UNLESS OTHERWISE STATED. Find the residual sum of squares, S_e .

• y is the observed value.

Step2

- $\dot{\mathbf{u}}$ is the the estimated value based on our regression equation.
- $y \hat{u}$ is called the residual and is written as e.

Note: The caret in $\hat{\mathbf{U}}$ is affectionately called a *hat*, so we call this parameter estimate *y*-hat.

	High		Predicted		
	temp.	Actual iced	iced tea		
	in °C	tea orders	orders	Residuals (e)	Squared residuals
	x	у	$\hat{y} = ax + b$	y – ŷ	$(\boldsymbol{y} - \hat{\boldsymbol{y}})^2$
22nd (Mon.)	29	77	$a \times 29 + b$	77 – $(a \times 29 + b)$	$[77 - (a \times 29 + b)]^2$
23rd (Tues.)	28	62	$a \times 28 + b$	$62 - (a \times 28 + b)$	$[62 - (a \times 28 + b)]^2$
24th (Wed.)	34	93	$a \times 34 + b$	93 – $(a \times 34 + b)$	$[93 - (a \times 34 + b)]^2$
25th (Thurs.)	31	84	$a \times 31 + b$	84 – $(a \times 31 + b)$	$[84 - (a \times 31 + b)]^2$
26th (Fri.)	25	59	$a \times 25 + b$	59 – $(a \times 25 + b)$	$[59 - (a imes 25 + b)]^2$
27th (Sat.)	29	64	$a \times 29 + b$	$64 - (a \times 29 + b)$	$[64 - (a \times 29 + b)]^2$
28th (Sun.)	32	80	$a \times 32 + b$	80 – $(a \times 32 + b)$	$[80 - (a \times 32 + b)]^2$
29th (Mon.)	31	75	$a \times 31 + b$	75 – $(a \times 31 + b)$	$[75 - (a \times 31 + b)]^2$
30th (Tues.)	24	58	$a \times 24 + b$	58 – $(a \times 24 + b)$	$[58 - (a \times 24 + b)]^2$
31st (Wed.)	33	91	$a \times 33 + b$	91 – $(a \times 33 + b)$	$[91 - (a \times 33 + b)]^2$
1st (Thurs.)	25	51	$a \times 25 + b$	51 - (a imes 25 + b)	$[51 - (a imes 25 + b)]^2$
2nd (Fri.)	31	73	$a \times 31 + b$	73 – $(a \times 31 + b)$	$[73 - (a \times 31 + b)]^2$
3rd (Sat.)	26	65	$a \times 26 + b$	$65 - (a \times 26 + b)$	$[65 - (a imes 26 + b)]^2$
4th (Sun.)	30	84	$a \times 30 + b$	84 – $(a \times 30 + b)$	$[84 - (a \times 30 + b)]^2$
Sum	408	1016	408a + 14b	1016 - (408a + 14b)	$S_e \blacktriangleleft$
Average	29.1	72.6	29.1a + b	72.6 - (29.1a + b)	S_e
			$=\overline{x}a+b$	$= \overline{y} - (\overline{x}a + b)$	$=\overline{14}$
	¥	¥			
	r	ū	$\mathbf{S}_e = [77 - (a \times 29)]$	$(b+b)$ ² + + $[84-(a \times 30+b)]$	$\left \right ^{2}$
THES	UM OF	THE RESID	MALS SQUA	REDIS	5-2 BT
CALLE	DTHE	RESIDUAL :	SUM OF SQ	VARES. >	

IT IS WRITTEN AS S_e OR RSS.

Differentiate
$$S_e$$
 with respect to a and b , and set it equal to 0.
When differentiating $y = (ax + b)^{n-1}$ with respect to x , the result is
 $\frac{dy}{dx} = n(ax + b)^{n-1} \times a$.
• Differentiate with respect to a .
 $\frac{dS_e}{da} = 2[77 - (29a + b)] \times (-29) + \dots + 2[84 - (30a + b)] \times (-30) = 0$

• Differentiate with respect to b.

$$\frac{dS_e}{db} = 2\left[77 - (29a + b)\right] \times (-1) + \dots + 2\left[84 - (30a + b)\right] \times (-1) = 0$$

Ø

a AND b.



Ster

Rearrange $\mathbf{0}$ and $\mathbf{0}$ from the previous step.

Rearrange **0**.

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\begin{split} & 2\Big[77-(29a+b)\Big]\times(-29)+\dots+2\Big[84-(30a+b)\Big]\times(-30)=0\\ & \Big[77-(29a+b)\Big]\times(-29)+\dots+\Big[84-(30a+b)\Big]\times(-30)=0\\ & 29\Big[(29a+b)-77\Big]+\dots+30\Big[(30a+b)-84\Big]=0\\ & (29\times29a+29\times b-29\times77)+\dots+(30\times30a+30\times b-30\times84)=0 \end{split} \label{eq:starseq} \qquad \text{MULTIPLY BY -1.}\\ & (29^2+\dots+30^2)a+(29+\dots+30)b-(29\times77+\dots+30\times84)=0 \end{aligned}
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Rearrange **2**.





Step6 Calculate the regression equation.

From **(c)** in Step 5,
$$a = \frac{S_{xy}}{S_{xx}}$$
. From **(c)** in Step 4, $b = \overline{y} - \overline{x}a$.

If we plug in the values we calculated in Step 1,

$$\begin{cases} a = \frac{S_{xx}}{S_{xy}} = \frac{484.9}{129.7} = 3.7\\ b = \bar{y} - \bar{x}a = 72.6 - 29.1 \times 3.7 = -36.4 \end{cases}$$

then the regression equation is

$$y = 3.7x - 36.4$$
.

It's that simple!

Note: The values shown are rounded for the sake of printing, but the result (36.4) was calculated using the full, unrounded values.





STEP 3: CALCULATE THE CORRELATION COEFFICIENT (R) AND ASSESS OUR POPULATION AND ASSUMPTIONS.







HERE'S THE EQUATION. WE CALCULATE THESE LIKE WE DID S_{xx} AND S_{xy} BEFORE.								
sum of products y and \hat{y} S _{un}								
$R = \frac{1}{\sqrt{\text{sum of squares of } y \times \text{sum of squares of } \hat{y}}} = \frac{gg}{\sqrt{S_{\text{m}} \times S_{\text{so}}}}$								
1812.3								
$=\frac{1012.0}{\sqrt{2203.4 \times 1812.3}}=0.9069$								
	V2203.4 × 1812.3 THAT'S NOT TOO BAD!							
	THIS LOOKS FAMILIAR. REGRESSION FUNCTION!							
	Actual	Estimated						
	values	$\hat{u} = 3.7x - 36.4$	11 – 11	$\hat{u} - \hat{\hat{u}}$	$(\boldsymbol{u}-\boldsymbol{\bar{u}})^2$	$(\hat{\boldsymbol{u}} - \bar{\hat{\boldsymbol{u}}})^2$	$(\boldsymbol{u}-\boldsymbol{\bar{u}})(\boldsymbol{\hat{u}}-\boldsymbol{\bar{\hat{u}}})$	$(\boldsymbol{u}-\boldsymbol{\hat{u}})^2$
22nd (Mon.)	77	72.0	4.4		19.6	0.3	-2.4	24.6
23rd (Tues.)	62	68.3	-10.6	-4.3	111.8	18.2	45.2	39.7
24th (Wed.)	93	90.7	20.4	18.2	417.3	329.6	370.9	5.2
25th (Thurs.)	84	79.5	11.4	6.9	130.6	48.2	79.3	20.1
26th (Fri.)	59	57.1	-13.6	-15.5	184.2	239.8	210.2	3.7
27th (Sat.)	64	72.0	-8.6	-0.5	73.5	0.3	4.6	64.6
28th (Sun.)	80	83.3	7.4	10.7	55.2	114.1	79.3	10.6
29th (Mon.)	75	79.5	2.4	6.9	5.9	48.2	16.9	20.4
30th (Tues.)	58	53.3	-14.6	-19.2	212.3	369.5	280.1	21.6
31st (Wed.)	91	87.0	18.4	14.4	339.6	207.9	265.7	16.1
1st (Thurs.)	51	57.1	-21.6	-15.5	465.3	239.8	334.0	37.0
2nd (Fri.)	73	79.5	0.4	6.9	0.2	48.2	3.0	42.4
3rd (Sat.)	65	60.8	-7.6	-11.7	57.3	138.0	88.9	17.4
4th (Sun.)	84	75.8	11.4	3.2	130.6	10.3	36.6	67.6
Sum	1016	1016	0	0	2203.4	1812.3	1812.3	391.1
Average	72.6	72.6	_					
	↓	¥			¥	¥	¥	¥
	ū	û			\mathbf{S}_{yy}	S _{ûû}	S _{iŷ}	\mathbf{S}_{e}
Se ISN'T NECESSARY FOR CALCULATING R, BUT I INCLUDED IT BECAUSE WE'LL NEED IT LATER.								





HERE, LOOK AT THE TEA ROOM DATA AGAIN.	22nd (Mon.) 23rd (Tues.) 24th (Wed.) 25th (Thurs.) 26th (Fri.) 27th (Sat.) 28th (Sun.) 29th (Mon.) 30th (Tues.) 31st (Wed.) 1st (Thurs.) 2nd (Fri.) 3rd (Sat.) 4th (Sun.)	High temp. (°C) 29 28 34 31 25 29 32 31 24 33 25 31 24 33 25 31 26 30	Iced tea orders 77 62 93 84 59 64 80 75 58 91 51 73 65 84	HOW MANY DAYS ARE THERE WITH A HIGH TEMPERATURE OF 31°C? THE 25TH, 29TH, AND 2ND SO THREE.
50			225 29 2 31°C	I CAN MAKE A CHART LIKE THIS FROM YOUR ANSWER.
NOW, CONSIDER THAT	THESE THRI ARE NOT TH DAYS IN HIS WITH A HIGH ARE THE 25th	EE DAYS E ONLY STORY OF 31°C, SY? 50 40 30 40 30 10 10 -10 -20 -30		THERE MUST HAVE BEEN MANY OTHERS IN THE PAST, AND THERE WILL BE MANY MORE IN THE FUTURE, RIGHT?









STEP 4: CONDUCT THE ANALYSIS OF VARIANCE.









THE STEPS OF ANOVA

Step 1	Define the population.	The population is "days with a high temperature of x degrees."
Step 2	Set up a null hypothesis and	Null hypothesis is $A = 0$.
	an alternative hypothesis.	Alternative hypothesis is $A \neq 0$.
Step 3	Select which hypothesis test to conduct.	We'll use analysis of one-way variance.
Step 4	Choose the significance level.	We'll use a significance level of .05.
Step 5	Calculate the test statistic	The test statistic is:
	from the sample data.	$\frac{a^2}{\left(\frac{1}{S_{xx}}\right)} \div \frac{S_e}{\text{number of individuals} - 2}$
		Plug in the values from our sample regression
		equation:
		$\frac{3.7^2}{\left(\frac{1}{129.7}\right)} \div \frac{391.1}{14-2} = 55.6$
		The test statistic will follow an F distribution
		with first degree of freedom 1 and second degree of
		freedom 12 (number of individuals minus 2), if the null hypothesis is true.
Step 6	Determine whether the <i>p</i> -value for the test statistic obtained in Step 5 is smaller than the significance level.	At significance level .05, with d_1 being 1 and d_2 being 12, the critical value is 4.7472. Our test statistic is 55.6.
Step 7	Decide whether you can reject	Since our test statistic is greater than the critical value,
	the null hypothesis.	we reject the null hypothesis.





SO A ≠ 0,

STEP 5: CALCULATE THE CONFIDENCE INTERVALS.











STEP 6: MAKE A PREDICTION!





IF YOU'RE DOING THE CALCULATION WITH THE FULL, UNROUNDED FIGURES, YOU SHOULD GET 64.6.









WHICH STEPS ARE NECESSARY?

Remember the regression analysis procedure introduced on page 68?

- 1. Draw a scatter plot of the independent variable versus the dependent variable. If the dots line up, the variables may be correlated.
- z. Calculate the regression equation.
- 3. Calculate the correlation coefficient (R) and assess our population and assumptions.
- 4. Conduct the analysis of variance.
- 5. Calculate the confidence intervals.
- 6. Make a prediction!

In this chapter, we walked through each of the six steps, but it isn't always necessary to do every step. Recall the example of Miu's age and height on page 25.

- · Fact: There is only one Miu in this world.
- Fact: Miu's height when she was 10 years old was 137.5 cm.

Given these two facts, it makes no sense to say that "Miu's height when she was 10 years old follows a normal distribution with mean Ax + B and standard deviation σ ." In other words, it's nonsense to analyze the population of Miu's heights at 10 years old. She was just one height, and we know what her height was.

In regression analysis, we either analyze the entire population or, much more commonly, analyze a sample of the larger population. When you analyze a sample, you should perform all the steps. However, since Steps 4 and 5 assess how well the sample represents the population, you can skip them if you're using data from an entire population instead of just a sample.

NOTE We use the term statistic to describe a measurement of a characteristic from a sample, like a sample mean, and parameter to describe a measurement that comes from a population, like a population mean or coefficient.

STANDARDIZED RESIDUAL

Remember that a *residual* is the difference between the *measured* value and the value *estimated* with the regression equation. The *standardized residual* is the residual divided by its estimated standard deviation. We use the standardized residual to assess whether a particular measurement deviates significantly from

the trend. For example, say a group of thirsty joggers stopped by Norns on the 4th, meaning that though iced tea orders were expected to be about 76 based on that day's high temperature, customers actually placed 84 orders for iced tea. Such an event would result in a large standardized residual.

Standardized residuals are calculated by dividing each residual by an estimate of its standard deviation, which is calculated using the residual sum of squares. The calculation is a little complicated, and most statistics software does it automatically, so we won't go into the details of the calculation here.

Table 2-1 shows the standardized residual for the Norns data used in this chapter.

	High temperature x	Measured number of orders of iced tea <i>y</i>	Estimated number of orders of iced tea $\hat{y} = 3.7x - 36.4$	Residual y-ŷ	Standardized residual
22nd (Mon.)	29	77	72.0	5.0	0.9
23rd (Tues.)	28	62	68.3	-6.3	-1.2
24th (Wed.)	34	93	90.7	2.3	0.5
25th (Thurs.)	31	84	79.5	4.5	0.8
26th (Fri.)	25	59	57.1	1.9	0.4
27th (Sat.)	29	64	72.0	-8.0	-1.5
28th (Sun.)	32	80	83.3	-3.3	-0.6
29th (Mon.)	31	75	79.5	-4.5	-0.8
30th (Tues.)	24	58	53.3	4.7	1.0
31st (Wed.)	33	91	87.0	4.0	0.8
1st (Thurs.)	25	51	57.1	-6.1	-1.2
2nd (Fri.)	31	73	79.5	-6.5	-1.2
3rd (Sat.)	26	65	60.8	4.2	0.8
4th (Sun.)	30	84	75.8	8.2	1.5

TABLE 2-1: CALCULATING THE STANDARDIZED RESIDUAL

As you can see, the standardized residual on the 4th is 1.5. If iced tea orders had been 76, as expected, the standardized residual would have been 0.

Sometimes a measured value can deviate so much from the trend that it adversely affects the analysis. If the standardized residual is greater than 3 or less than -3, the measurement is considered an *outlier*. There are a number of ways to handle outliers, including removing them, changing them to a set value, or just keeping them in the analysis as is. To determine which approach is most appropriate, investigate the underlying cause of the outliers.

INTERPOLATION AND EXTRAPOLATION

If you look at the x values (high temperature) on page 64, you can see that the highest value is 34° C and the lowest value is 24° C. Using regression analysis, you can *interpolate* the number of iced tea orders on days with a high temperature between 24° C and 34° C and *extrapolate* the number of iced tea orders on days with a high below 24° C or above 34° C. In other words, extrapolation is the estimation of values that fall outside the range of your observed data.

Since we've only observed the trend between 24°C and 34°C, we don't know whether iced tea sales follow the same trend when the weather is extremely cold or extremely hot. Extrapolation is therefore less reliable than interpolation, and some statisticians avoid it entirely.

For everyday use, it's fine to extrapolate—as long as you're aware that your result isn't completely trustworthy. However, avoid using extrapolation in academic research or to estimate a value that's far beyond the scope of the measured data.

AUTOCORRELATION

The independent variable used in this chapter was high temperature; this is used to predict iced tea sales. In most places, it's unlikely that the high temperature will be 20°C one day and then shoot up to 30°C the next day. Normally, the temperature rises or drops gradually over a period of several days, so if the two variables are related, the number of iced tea orders should rise or drop gradually as well. Our assumption, however, has been that the deviation (error) values are random. Therefore, our predicted values do not change from day to day as smoothly as they might in real life.

When analyzing variables that may be affected by the passage of time, it's a good idea to check for autocorrelation. Autocorrelation occurs when the error is correlated over time, and it can indicate that you need to use a different type of regression model.

There's an index to describe autocorrelation—the Durbin-Watson statistic, which is calculated as follows:

$$d = \frac{\sum_{t=2}^{T} (e_t - e_{t-1})^2}{\sum_{t=1}^{T} e_t^2}$$

The equation can be read as "the sum of the square of each residual minus the previous residual, divided by the sum of each residual squared." You can calculate the value of the Durbin-Watson statistic for the example in this chapter:

$$\frac{(-6.3-5.0)^2+(2.3-(-6.3))^2+\dots+(8.2-4.2)^2}{5.0^2+(-6.3)^2+\dots+8.2^2}=1.8$$

The exact critical value of the Durbin-Watson test differs for each analysis, and you can use a table to find it, but generally we use 1 as a cutoff: a result less than 1 may indicate the presence of autocorrelation. This result is close to 2, so we can conclude that there is no autocorrelation in our example.

NONLINEAR REGRESSION

On page 66, Risa said:



This equation is linear, but regression equations don't have to be linear. For example, these equations may also be used as regression equations:

$$y = \frac{a}{r} + b$$

$$\cdot \quad \mathbf{u} = \mathbf{a}\sqrt{\mathbf{x}} + \mathbf{b}$$

$$\cdot y = ax^2 + bx + c$$

 $\cdot y = a \times \log x + b$

The regression equation for Miu's age and height introduced on page 26 is actually in the form of $y = \frac{a}{x} + b$ rather than y = ax + b.

Of course, this raises the question of which type of equation you should choose when performing regression analysis on your own data. Below are some steps that can help you decide.

- Draw a scatter plot of the data points, with the dependent variable values on the x-axis and the independent variable values on the y-axis. Examine the relationship between the variables suggested by the spread of the dots: Are they in roughly a straight line? Do they fall along a curve? If the latter, what is the shape of the curve?
- 2. Try the regression equation suggested by the shape in the variables plotted in Step 1. Plot the residuals (or standardized residuals) on the y-axis and the independent variable on the x-axis. The residuals should appear to be random, so if there is an obvious pattern in the residuals, like a curved shape, this suggests that the regression equation doesn't match the shape of the relationship.
- 3. If the residuals plot from Step 2 shows a pattern in the residuals, try a different regression equation and repeat Step 2. Try the shapes of several regression equations and pick one that appears to most closely match the data. It's usually best to pick the simplest equation that fits the data well.

TRANSFORMING NONLINEAR EQUATIONS INTO LINEAR EQUATIONS

There's another way to deal with nonlinear equations: simply turn them into linear equations. For an example, look at the equation for Miu's age and height (from page 26):

$$y = -\frac{326.6}{x} + 173.3$$

You can turn this into a linear equation. Remember:

If
$$\frac{1}{x} = X$$
, then $\frac{1}{X} = x$.

So we'll define a new variable X, set it equal to $\frac{1}{x}$, and use X in the normal y = aX + b regression equation. As shown on page 76, the value of a and b in the regression equation y = aX + b can be calculated as follows:

$$\begin{cases} \boldsymbol{a} = \frac{\mathbf{S}_{Xy}}{\mathbf{S}_{XX}} \\ \boldsymbol{b} = \overline{\boldsymbol{y}} - \overline{X}\boldsymbol{a} \end{cases}$$

		1 age						
	Age	1	Height					
	x	$\frac{\mathbf{I}}{\mathbf{x}} = \mathbf{X}$	y	$\left(oldsymbol{x} - \overline{oldsymbol{x}} ight)$	$oldsymbol{y} - oldsymbol{\overline{y}}$	$\left(\boldsymbol{X} - \overline{\boldsymbol{X}} \right)^2$	$(\boldsymbol{y}-\bar{\boldsymbol{y}})^2$	$ig(oldsymbol{X} - \overline{oldsymbol{X}} ig) ig(oldsymbol{y} - \overline{oldsymbol{y}} ig)$
	4	0.2500	100.1	0.1428	-38.1625	0.0204	1456.3764	-5.4515
	5	0.2000	107.2	0.0928	-31.0625	0.0086	964.8789	-2.8841
	6	0.1667	114.1	0.0595	-24.1625	0.0035	583.8264	-1.4381
	7	0.1429	121.7	0.0357	-16.5625	0.0013	274.3164	-0.5914
	8	0.1250	126.8	0.0178	-11.4625	0.0003	131.3889	-0.2046
	9	0.1111	130.9	0.0040	-7.3625	0.0000	54.2064	-0.0292
	10	0.1000	137.5	-0.0072	-0.7625	0.0001	0.5814	-0.0055
	11	0.0909	143.2	-0.0162	4.9375	0.0003	24.3789	-0.0802
	12	0.0833	149.4	-0.0238	11.1375	0.0006	124.0439	-0.2653
	13	0.0769	151.6	-0.0302	13.3375	0.0009	177.889	-0.4032
	14	0.0714	154.0	-0.0357	15.7375	0.0013	247.6689	-0.5622
	15	0.0667	154.6	-0.0405	16.3375	0.0016	266.9139	-0.6614
	16	0.0625	155.0	-0.0447	16.7375	0.0020	280.1439	-0.7473
	17	0.0588	155.1	-0.0483	16.8375	0.0023	283.5014	-0.8137
	18	0.0556	155.3	-0.0516	17.0375	0.0027	290.2764	-0.8790
	19	0.0526	155.7	-0.0545	17.4375	0.0030	304.0664	-0.9507
Sum	184	1.7144	2212.2	0.0000	0.0000	0.0489	5464.4575	-15.9563
Average	11.5	0.1072	138.3					

TALBE 2-2: CALCULATING THE REGRESSION EQUATION

According to the table:

$$\begin{cases} a = \frac{S_{xy}}{S_{xx}} = \frac{-15.9563}{0.0489} = -326.6^* \\ b = \overline{y} - \overline{X}a = 138.2625 - 0.1072 \times (-326.6) = 173.3 \end{cases}$$

So the regression equation is this:

$$y = -326.6X + 173.3$$

$$\uparrow \qquad \uparrow$$
height
$$\frac{1}{age}$$

^{*} If your result is slightly different from 326.6, the difference might be due to rounding. If so, it should be very small.

which is the same as this:

$$y = -\frac{326.6}{x} + 173.3$$

$$\uparrow \qquad \uparrow$$
height age

We've transformed our original, nonlinear equation into a linear one!