

## Unit 5: Regression

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## Fact

- *The scatter plot generally clusters about the SD line, but does not, in general, give the line of linear correlation.*
- *It goes through the ends of the football, so gives the direction of the football.*
- *Not used for predictions, only for location purposes on a scatterplot.*
- *Goes through  $(\text{Avg}_x, \text{Avg}_y)$  with slope  $\pm SD_y / SD_x$  depending on  $r > 0$  or  $r < 0$ .*
- *Hence, the line is*

$$y = \pm \frac{SD_y}{SD_x} (x - \text{Avg}_x) + \text{Avg}_y.$$

## Fact

- *Has nothing to do with  $r$  (aside from  $r < 0$  or  $r > 0$ ), so we can have same SD-line for sets of data with identical averages, SDs, regardless of respective  $r$  values.*
- *Clearly, this is not what we want.*

## Fact

**Regression**=Attempt to quantify how one variable depends on another given a certain degree of correlation.

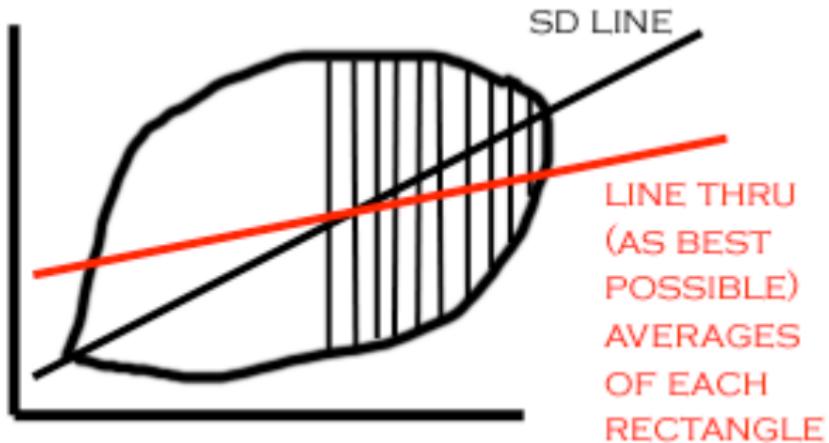
## Example

### Example

Suppose that we measure the relationship between height and weight. Let  $X$  =height,  $Y$  =weight. We have  $\text{Avg}_x = 70$  in.,  $\text{Avg}_y = 162$  lbs.,  $SD_x = 3$  in.,  $SD_y = 30$  lbs., with  $r = .47$ .

- Q: So if we know someone is 73 inches tall, what we predict for their weight?
- A: **Average** weight of all 73 inch tall people in the study.
- How do we get this?

# The regression line (best fit line)



## Fact

*The line in red is called the regression line for predicting  $y$  given  $x$ .*

## Fact

- *If the data is nice (normally distributed and linear correlated), then the average of  $y$  on each rectangle will be on a line.*
- *Graphically the regression line goes through the Avg point **and** the sides of the football.*

## Definition

- The regression line is given by the formula

$$y - \text{Avg}_y = r \frac{SD_y}{SD_x} (x - \text{Avg}_x).$$

- Note that the variable being predicted is on the left.
- By construction it goes through the point  $(\text{Avg}_x, \text{Avg}_y)$ .

## Fact

- We can rewrite the regression line as

$$\frac{y - \text{Avg}_y}{SD_y} = r \frac{x - \text{Avg}_x}{SD_x}.$$

- Thus, in standard units,

$$Y = rX.$$

# Example

## Example (Weight vs Height, cont'd)

What is the average weight of those 73" tall?

## Another example

### Example

Suppose that Colgate students who take the LSAT have  $\text{Avg}_{\text{LSAT}} = 600$ ,  $SD_{\text{LSAT}} = 50$ ,  $\text{Avg}_{\text{GPA}} = 3.3$ ,  $SD_{\text{GPA}} = 0.3$  and the correlation coefficient  $r = 0.6$ . What is the regression prediction for the LSAT score of the students with GPA of

- 3.3?
- 3.6?

# The regression fallacy

## Fact

- *A predicted value of  $y$  will be close to the average of  $y$  (in terms of  $SDs$ ) than the  $x$  value is to the average of  $x$  (in terms of  $SDs$ ).*
- *In virtually all test-retest situations, the bottom group's score will, **on average**, improve and the top group's scores will, **on average**, decline.*

## Example

### Example (Are sophomores dumber?)

Suppose that the average GPA for the first and second year students at Colgate is 2.8, SD is 0.7, and the correlation coefficient is  $r = 0.8$ . What is the most likely outcome for the best student (4.0 GPA as a first year student) in the second year?

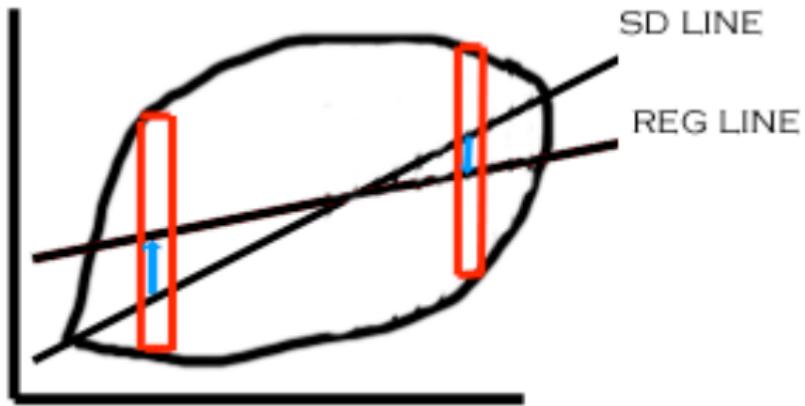
# The regression fallacy

## Fact

*Remember:*

- *Regression is an idealized description of a population.*
- *It doesn't describe individual cases.*
- *The previous example does not say that sophomores do worse than in the first year.*
- *It says that the best 1<sup>st</sup> years will on average do worse in their 2<sup>nd</sup> year, even though the overall average and SD stay the same.*

# The regression fallacy



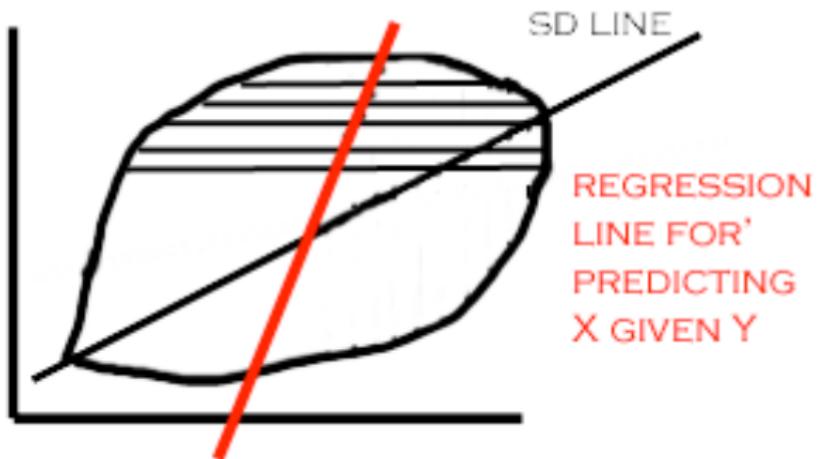
## Fact

*Blue arrows show increase in scores for those below average and decrease in scores for those above average.*

# There are TWO regression lines

## Fact

*One for predicting  $y$  given  $x$ ; one for predicting  $x$  given  $y$ .*



## Example

### Example

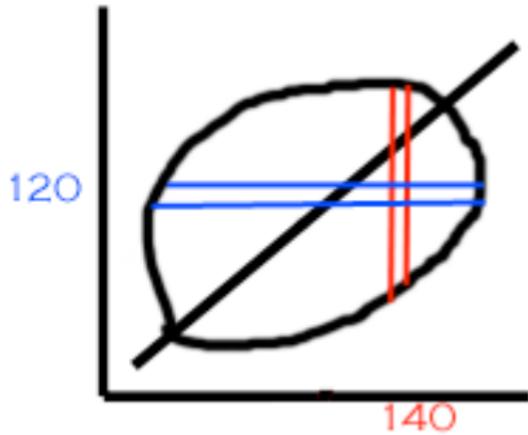
Suppose that the average IQ is 100 and SD is 15 for both women and men and husband/wife couples have a correlation coefficient of 0.5.

- What is the predicted average IQ of women whose husbands have IQ of 140?
- What do you predict for the average IQ of their husbands?

### Fact

- *So smart husbands have dumber wives, who in turn have dumber husbands.*
- *What gives?*
- *Again, different families of data under consideration:*
  - ① *All wives whose husbands have IQs of 140;*
  - ② *All husbands whose wives have IQs of 120:*

## Example (cont'd)

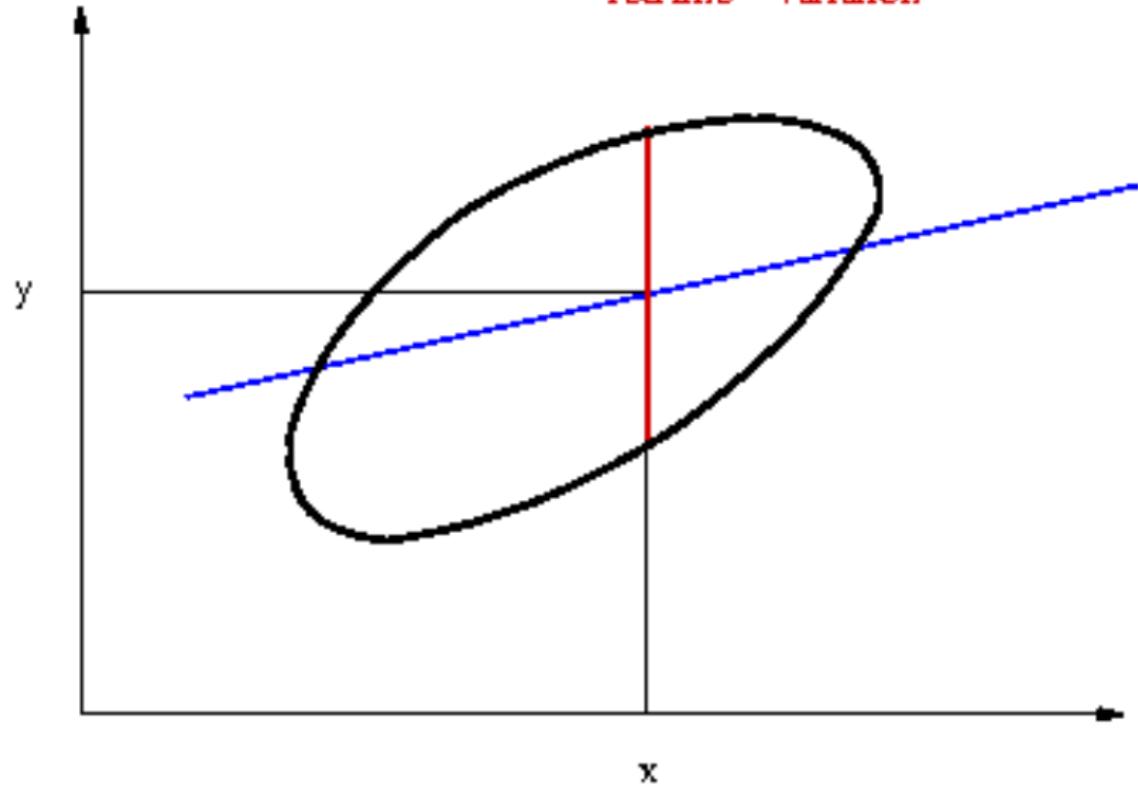


## Fact

- *Recall that if  $x$  and  $y$  are linearly correlated and  $x = x_1$  then the value  $y$  predicted using the regression line is the **average value***
- *What is the variation expected around this?*

# Variation

red line = variation



# The RMS error

## Definition

The RMS error is

$\sqrt{\text{Average of the squared errors.}}$

## Fact

- *The regression line is the line which minimizes the RMS error.*
- *For this reason the regression line is called also the **least square error** or **least squares** line.*

## Fact

- *The RMS error is a vertical measure of the error.*
- *For ideal situation, 68% of data will fall within 1 RMS error of the regression line.*
- *95% of data will be within 2 RMS errors of the regression line.*
- *The RMS error behaves like the SD from the regression line.*

# Formula for computing the RMS

## Fact

- *We can compute the RMS using the formula*

$$\text{RMS error} = \sqrt{1 - r^2} \cdot SD_y.$$

- *This formula is valid for normally distributed linearly correlated data.*

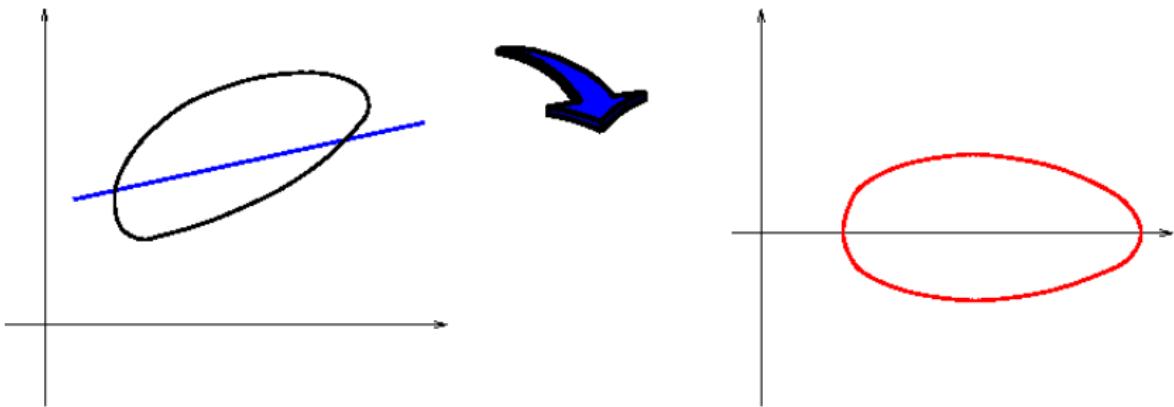
## Example

### Example

Suppose that for Colgate freshmen the average of the SAT verbal is 550, SD is 50, the average of their GPA is 2.8, SD is 0.7, and  $r = 0.5$ .

- What is the RMS error when predicting the SAT verbal?
- For students scoring 650 on the SAT verbal, what is the predicted GPA with error estimate? .

# Residuals



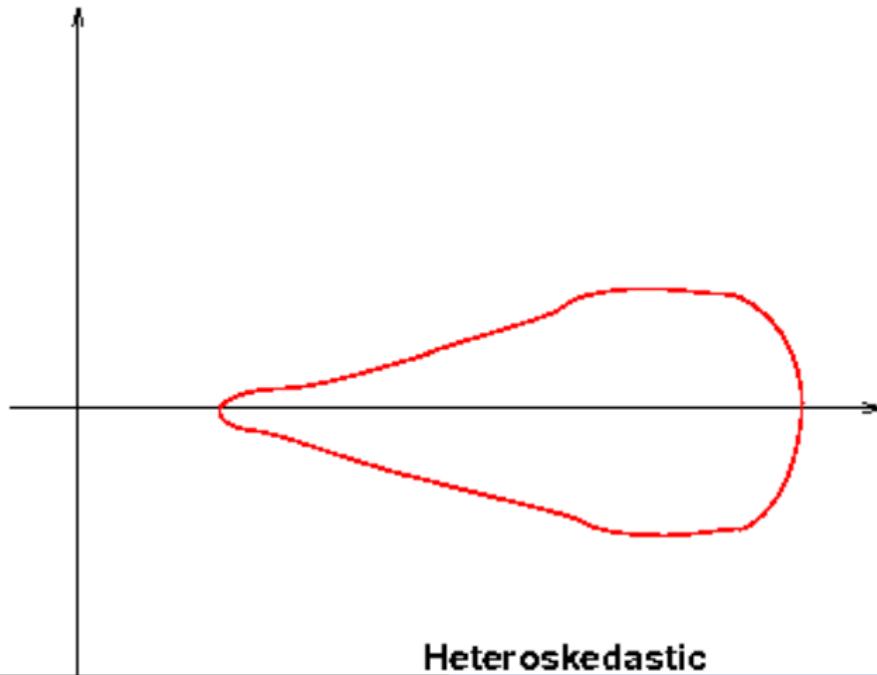
## Fact

- *The residuals are the errors centered around 0.*
- *They should have **no** “linear trend” (that is, the slope should be 0).*

# Heteroskedastic data

## Definition

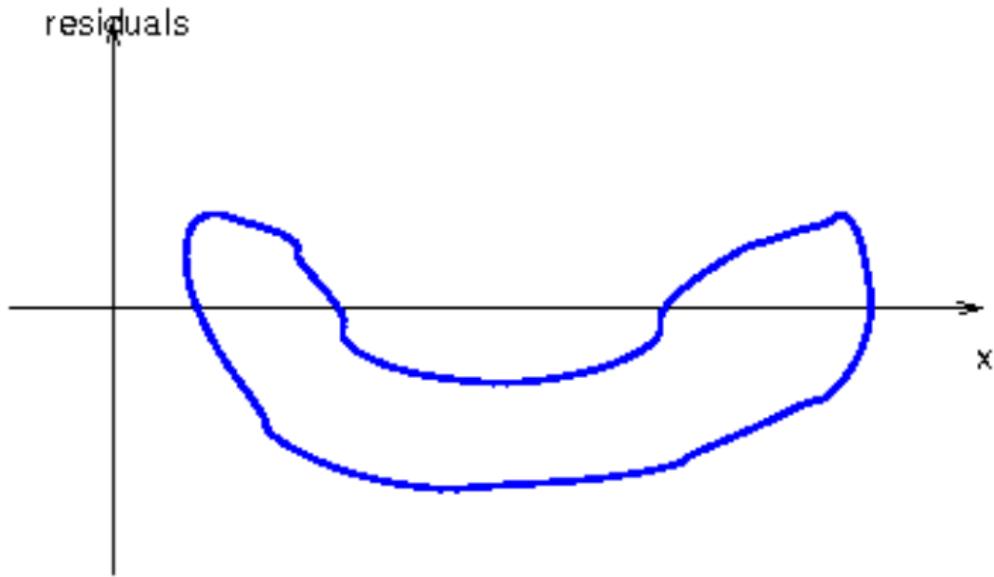
We say that data is heteroskedastic if the residuals have different distributions



## Fact

*RMS error formula is not a good error estimate for heteroskedastic data.*

# Nonlinear relationship



**Nonlinear relationship**

- Don't use the regression line!