

# Multilinear and Nonlinear Regression

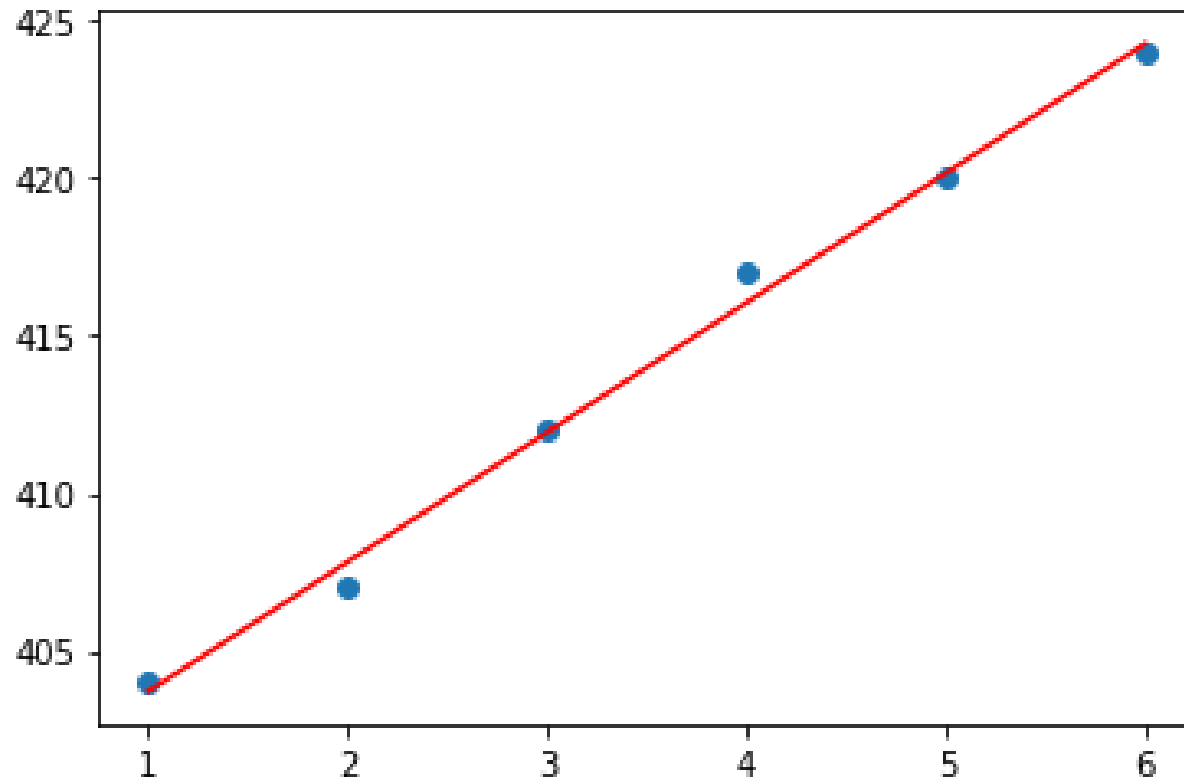
## Lecture 6c – 2021-06-18

MAT A35 – Summer 2021 – UTSC

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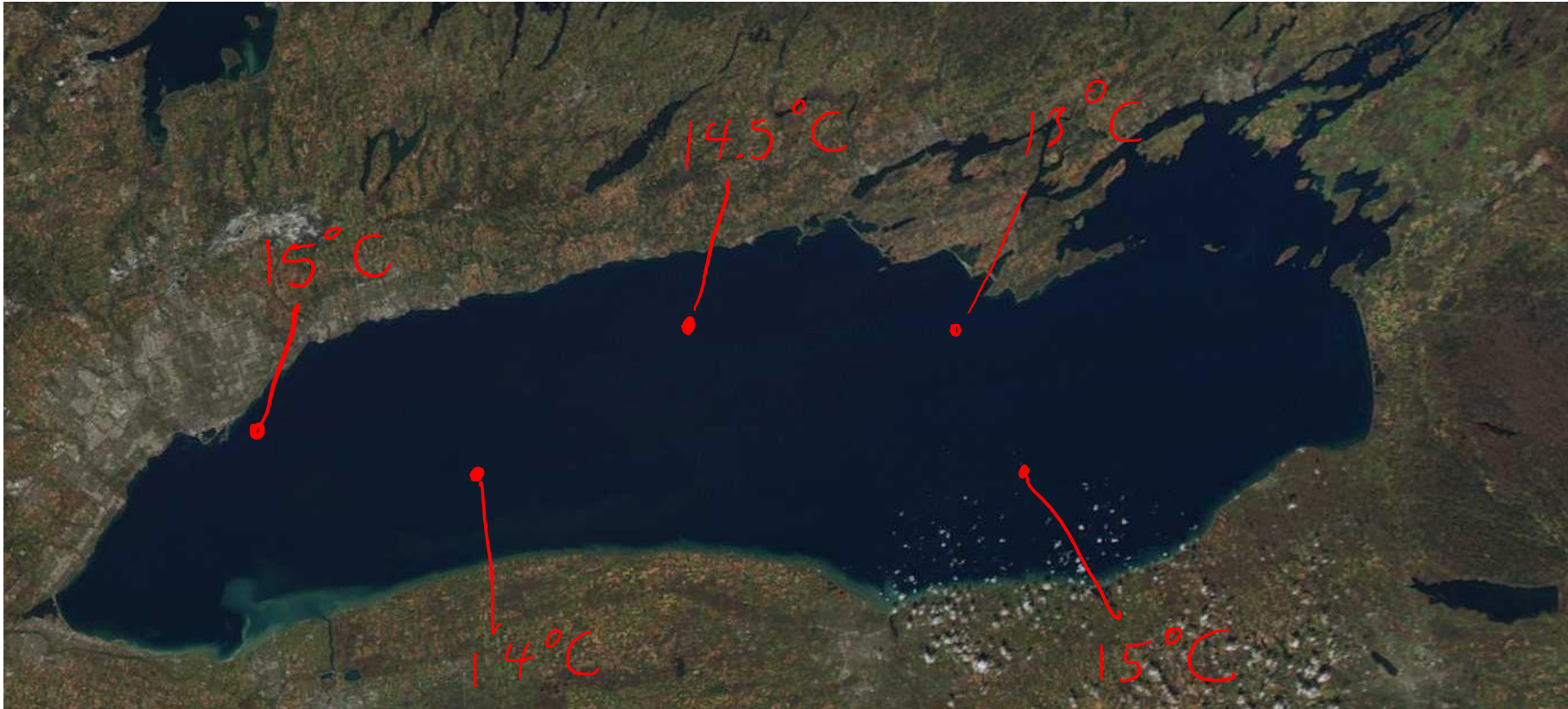
# Single variable linear regression

- Given samples of the dependent variable  $y_1, \dots, y_n$  at values of the independent variable  $x_1, \dots, x_n$ , we want to find the linear model  $f(x) = mx + b$  such that  $y_i \approx f(x_i)$ , the “best-fit” line.



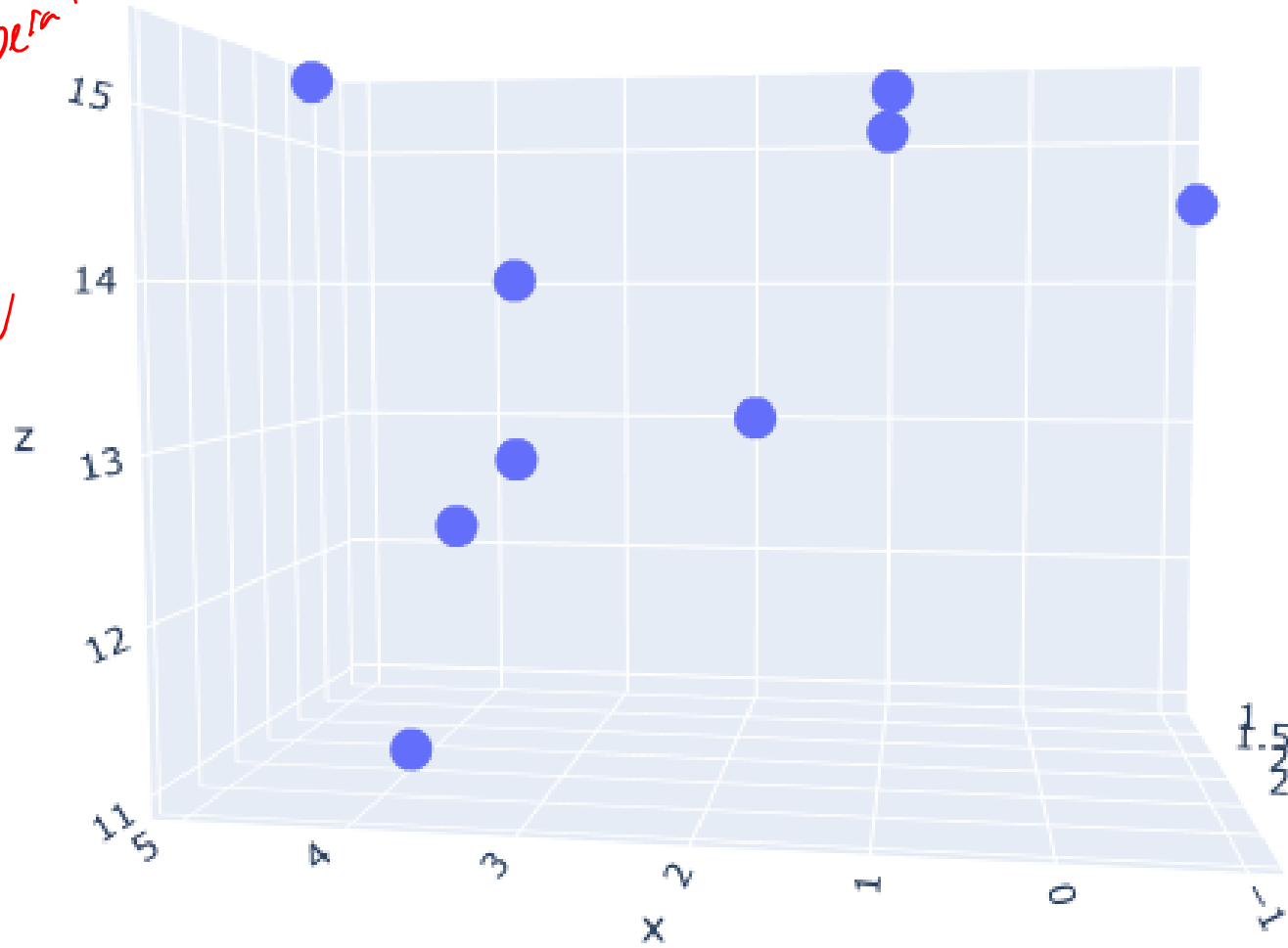
# Two-variable linear regression

- What if we have multiple independent variables?
- Suppose we are measuring the water temperature in Lake Ontario, and want to know how the temperature varies as a function of location

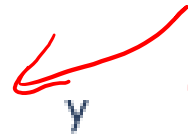


# 3D Scatter Plot of temperatures

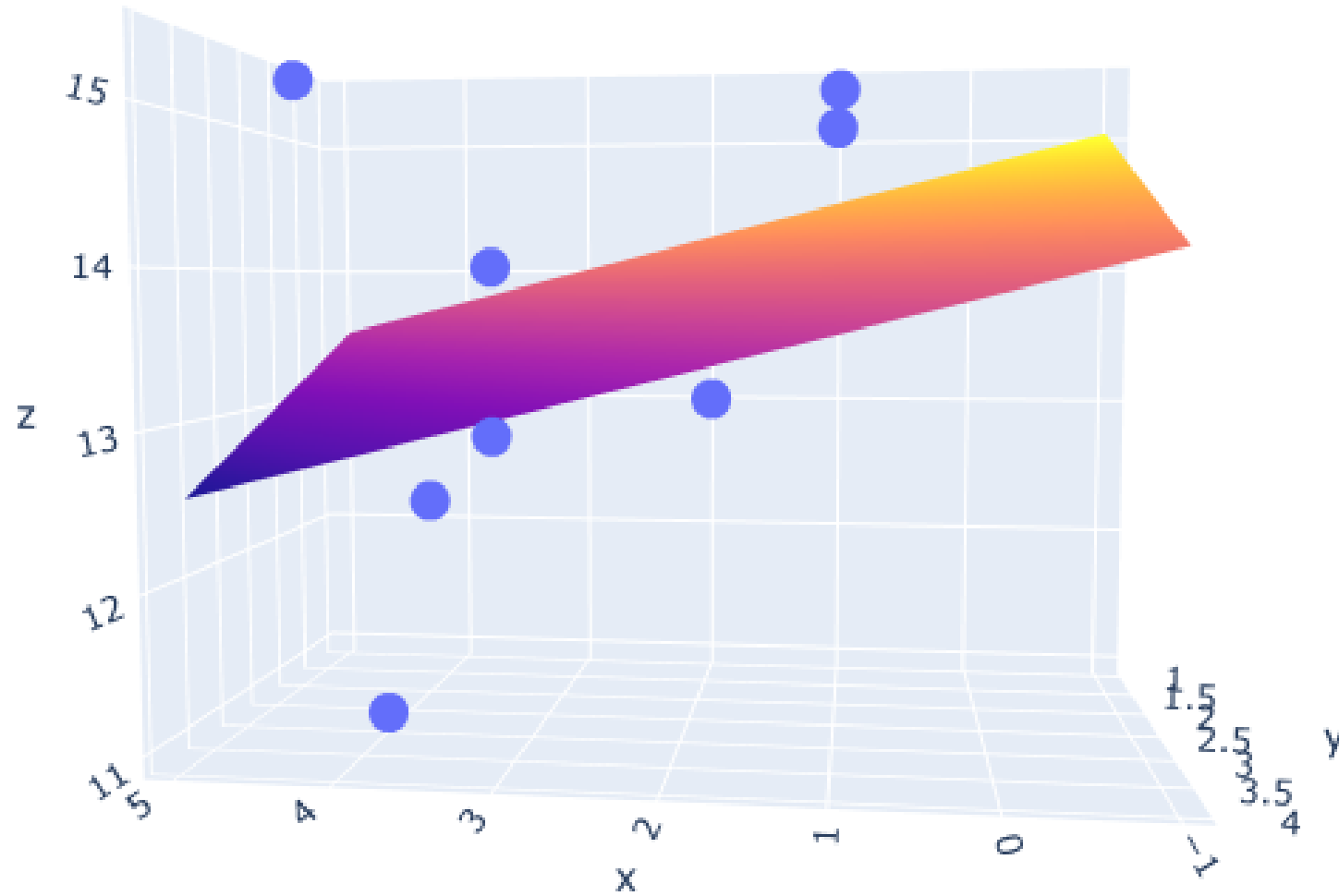
temperature



position  
in  
Lake  
Ontario



# Best-fit plane

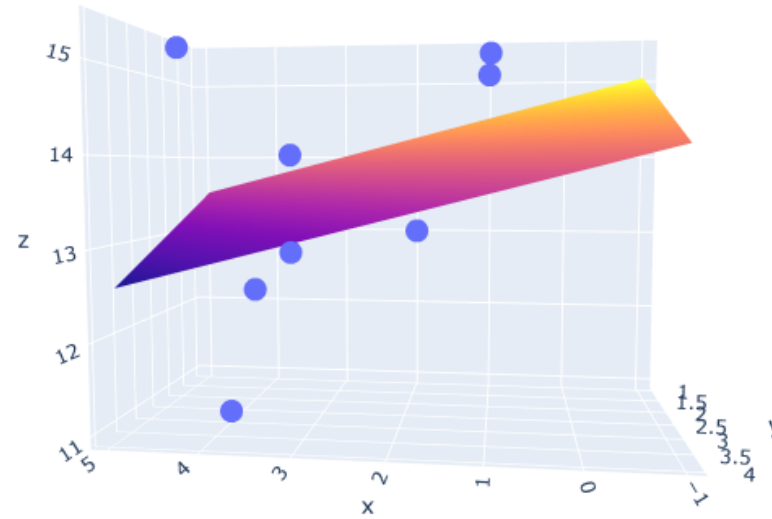
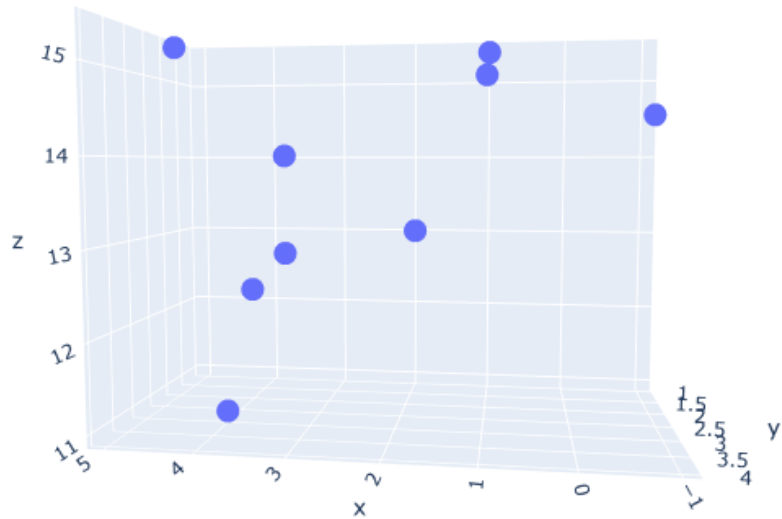


# Two-variable linear regression

- Let  $x$  and  $y$  be the independent variables. Let  $z$  be the dependent variable. Given samples  $z_1, \dots, z_n$  at values  $(x_1, y_1), \dots, (x_n, y_n)$ , we want the linear model

$$f(x, y) = m_1x + m_2y + b$$

such that  $z_i \approx f(x_i, y_i)$ , the “best-fit” plane.



# Multilinear regression

- One independent variable, one dependent variable

Model:  $f(x) = mx + b$

- Two independent variables, one dependent variable

Model:  $f(x, y) = m_1 x + m_2 y + b$   $f: \mathbb{R}^2 \rightarrow \mathbb{R}$

$\Leftrightarrow f\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = [m_1 \ m_2] \begin{bmatrix} x \\ y \end{bmatrix} + b$

- Many independent variables, one dependent variable

Model:  $f\left(\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}\right) = [m_1 \ \dots \ m_n] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} + b$   $f: \mathbb{R}^n \rightarrow \mathbb{R}$

$f: \mathbb{R}^n \rightarrow \mathbb{R}^p$

- Can also have many independent variables, many dependent...

# Try it out

- You are measuring the temperature of Lake Ontario as a function of location. You get the following data:

Longitude	Latitude	Temperature
76.5 W	43.5 N	12.2
76.5 W	43.9 N	12.1
77.0 W	43.6 N	11.6
77.0 W	43.8 N	11.5
78.0 W	43.3 N	13.7
78.0 W	43.7 N	13.1
79.5 W	43.8 N	12.3
79.5 W	43.9 N	12.1

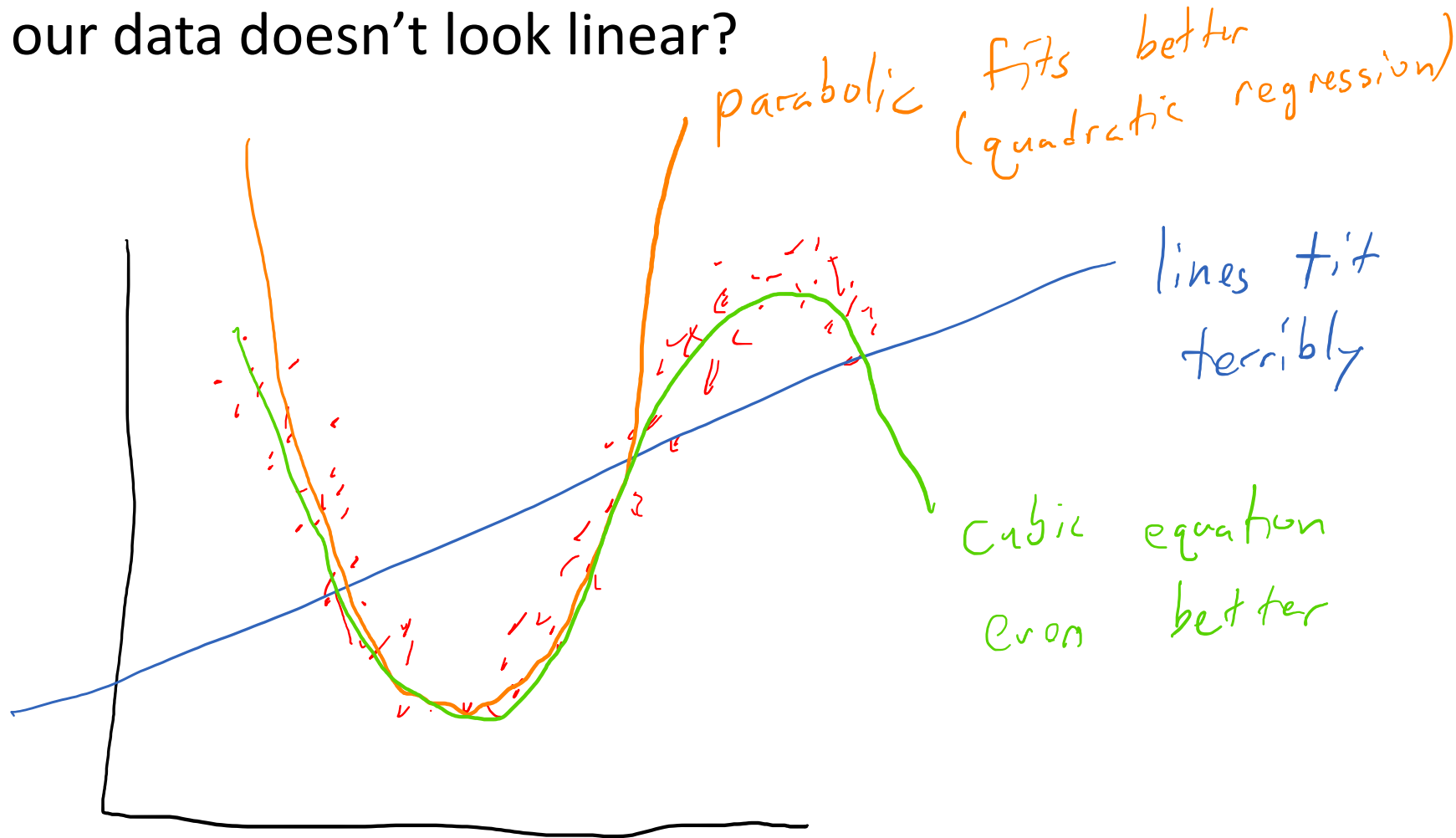
- A: 12.06
- B: 12.35
- C: 12.54
- D: 12.89
- E: None of the above

- The GPS coordinates for the lake near Toronto are 43.6 N, 79.3 W. What do you predict the lake water temperature to be near Toronto?



# Nonlinear regression

- What if our data doesn't look linear?

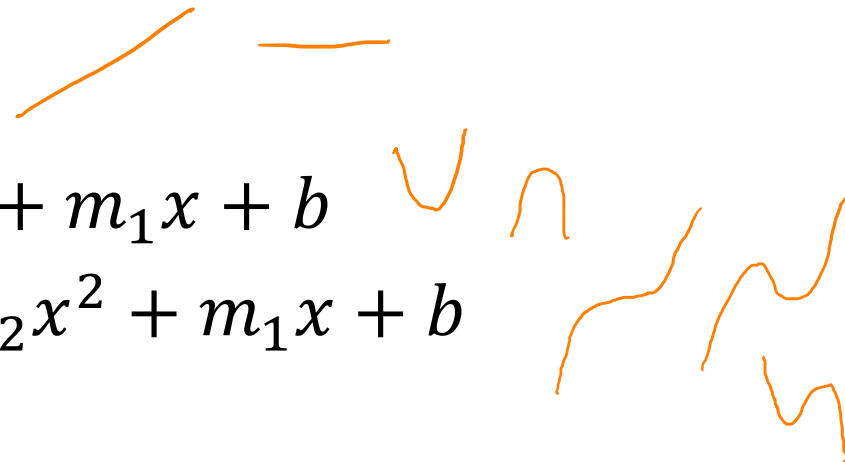


# Different types of regression

- Linear regression:  $f(x) = mx + b$
- Quadratic regression:  $f(x) = m_2x^2 + m_1x + b$
- Cubic regression:  $f(x) = m_3x^3 + m_2x^2 + m_1x + b$
- Polynomial regression of degree  $n$ :

$$f(x) = b + \sum_{i=1}^n m_i x^i$$

- Exponential regression:  $f(x) = c_1 e^{c_2 x}$
- Power dependencies:  $f(x) = c_1 x^{c_2}$



# Convert nonlinear to multilinear

Quadratic:  $f(x) = m_2 x^2 + m_1 x + b$

Let  $y = x^2$   $f(x, y) = m_2 y + m_1 x + b$

x	$y = x^2$	$z = x^3$	f(x)
1	1	1	0.1
2	4	8	5.2
-1	1	-1	6.9
4	16	64	4.2

Cubic:  $f(x) = m_3 x^3 + m_2 x^2 + m_1 x + b$

Let  $x = x$ ,  $y = x^2$ ,  $z = x^3$

$$f\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = [m_1 \ m_2 \ m_3] \begin{bmatrix} x \\ y \\ z \end{bmatrix} + b$$

# Intuition guess

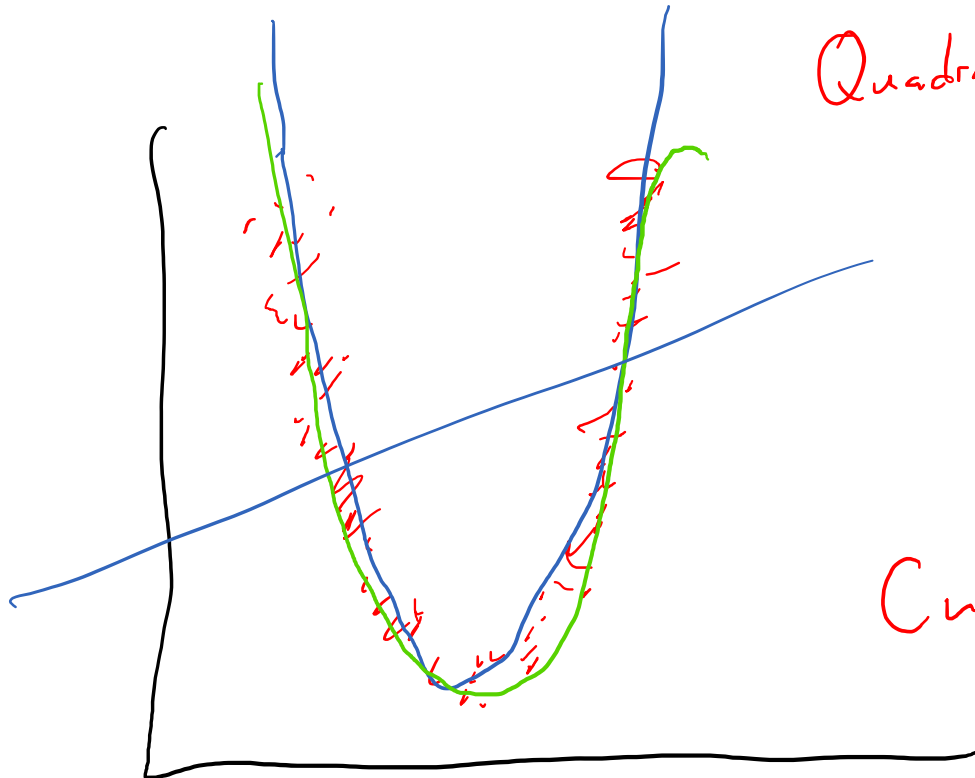
- Linear vs. Quadratic vs Cubic: which model will have smaller Mean Square Error for the following data:

Quadratic:  $f(x) = m_2 x^2 + m_1 x + b$

Cubic:  $f(x) = m_3 x^3 + m_2 x^2 + m_1 x + b$   
What if  $m_3 = 0$ ?

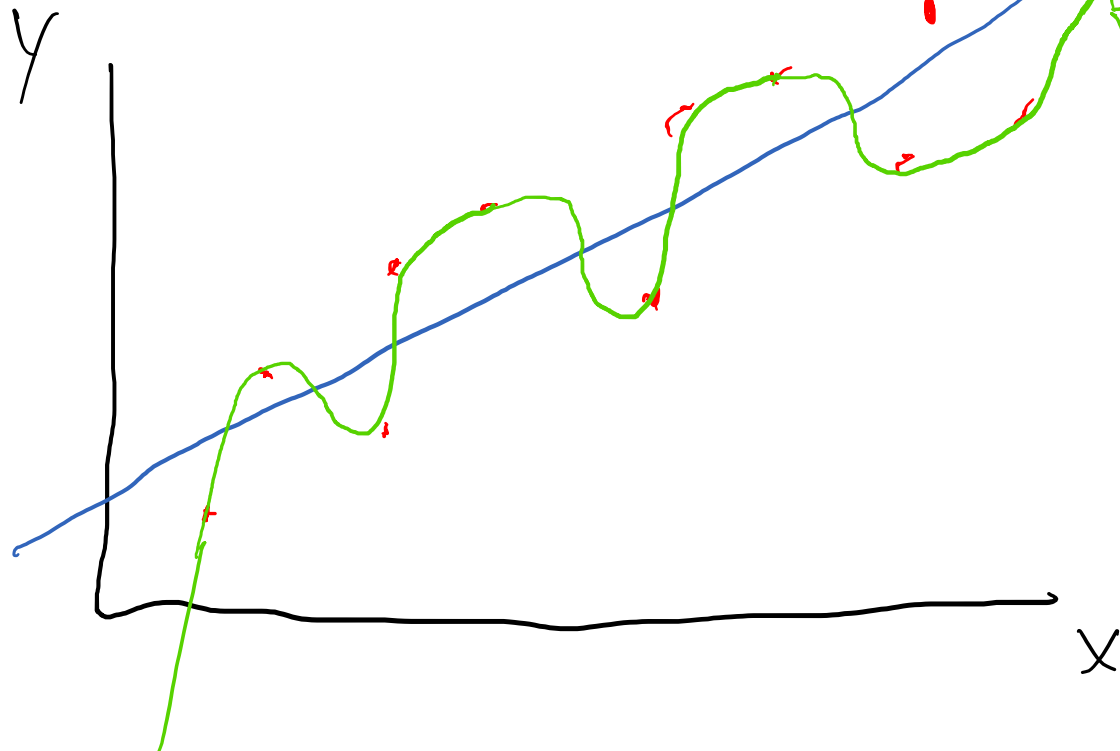
- A: Linear
- B: Quadratic
- C: Cubic
- D: Same error for All
- E: None of the above

For the same data, cubic will always have no higher error than quadratic.



# Be careful about too many parameters

- The more parameters you have (e.g. in a polynomial regression), the better your mean squared error will be.
- However, sometimes, you will overfit to the data.
- John von Neumann: “with four parameters, I can fit an elephant, and with five I can make him wiggle his trunk”



← better fit,  
but using more  
parameters.

OVERFIT

# Exponential regression

Exponential:

$$f(x) = c_1 e^{c_2 x}$$

$$\begin{aligned} \ln f(x) &= \ln c_1 e^{c_2 x} = \ln c_1 + \ln e^{c_2 x} \\ &= \ln c_1 + c_2 x \end{aligned}$$

Let  $z = \ln f(x)$ ,  $m = c_2$ ,  $b = \ln c_1$   
 $\Rightarrow z = mx + b$   $(f(x) = e^z, c_2 = m, c_1 = e^b)$

$x$	$f(x)$
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$x$	$\ln f(x)$
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# Power dependencies

$$f(x) = c_1 x^{c_2}$$

$$\ln f(x) = \ln c_1 + \ln x^{c_2} = \ln c_1 + c_2 \ln x$$

Let  $z = \ln f(x)$ ,  $m = c_2$ ,  $y = \ln x$ ,  $b = \ln c_1$   
 $f(x) = e^z$ ,  $x = e^y$ ,  $c_1 = e^b$

$$z = my + b \quad \leftarrow \text{linear regression problem}$$

ind.	dep.
$x$	$f(x)$

ind.	dep.
$y = \ln x$	$z = \ln f(x)$

↳ Log Log plots