

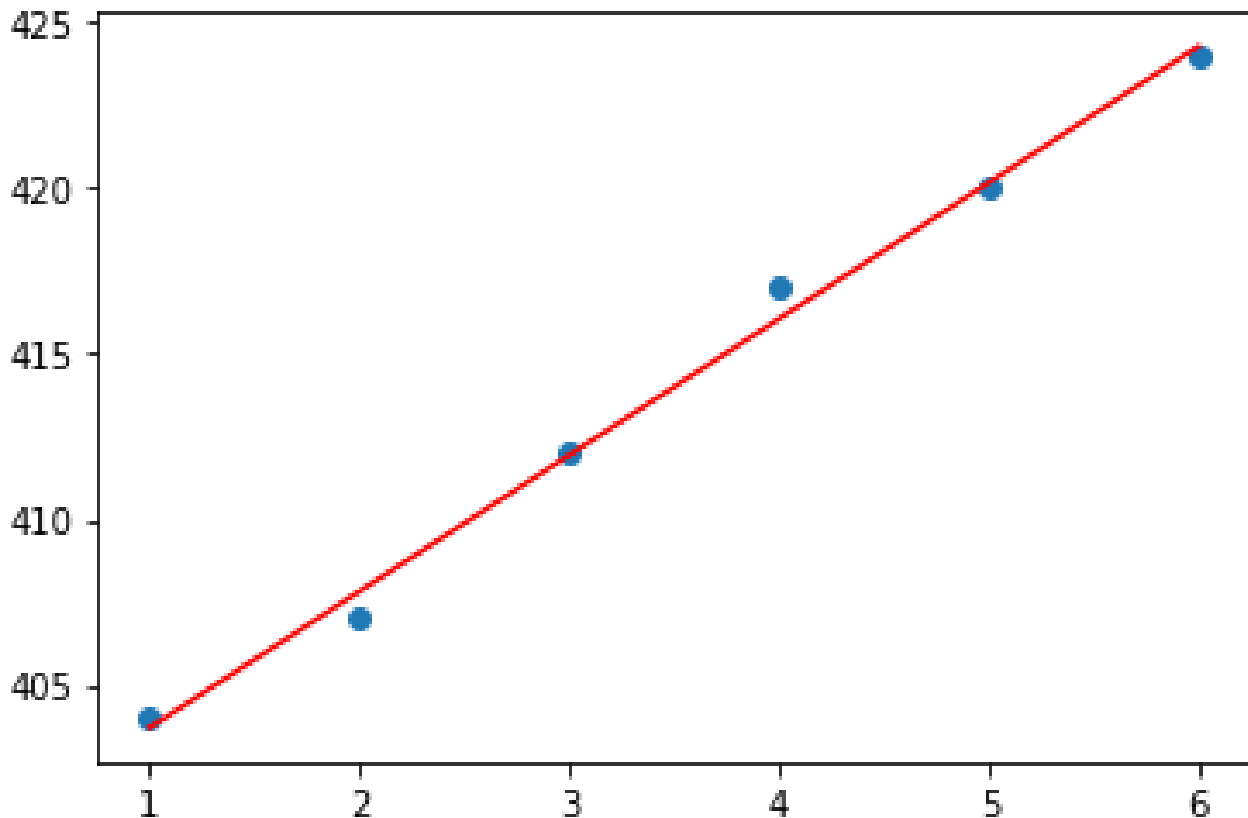
# Multilinear and Nonlinear Regression Lecture 6c: 2023-02-16

MAT A35 – Winter 2023 – UTSC

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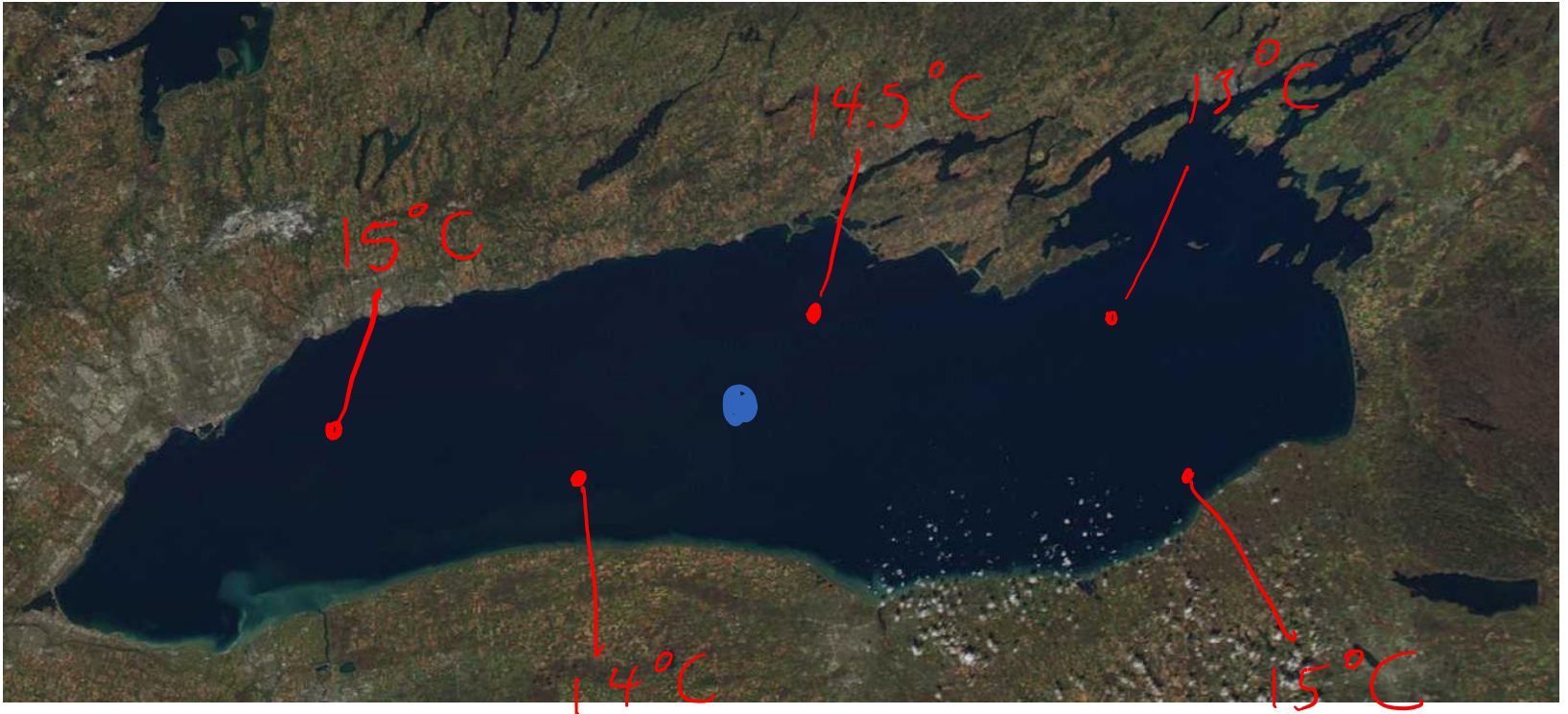
# Single variable linear regression

- 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.

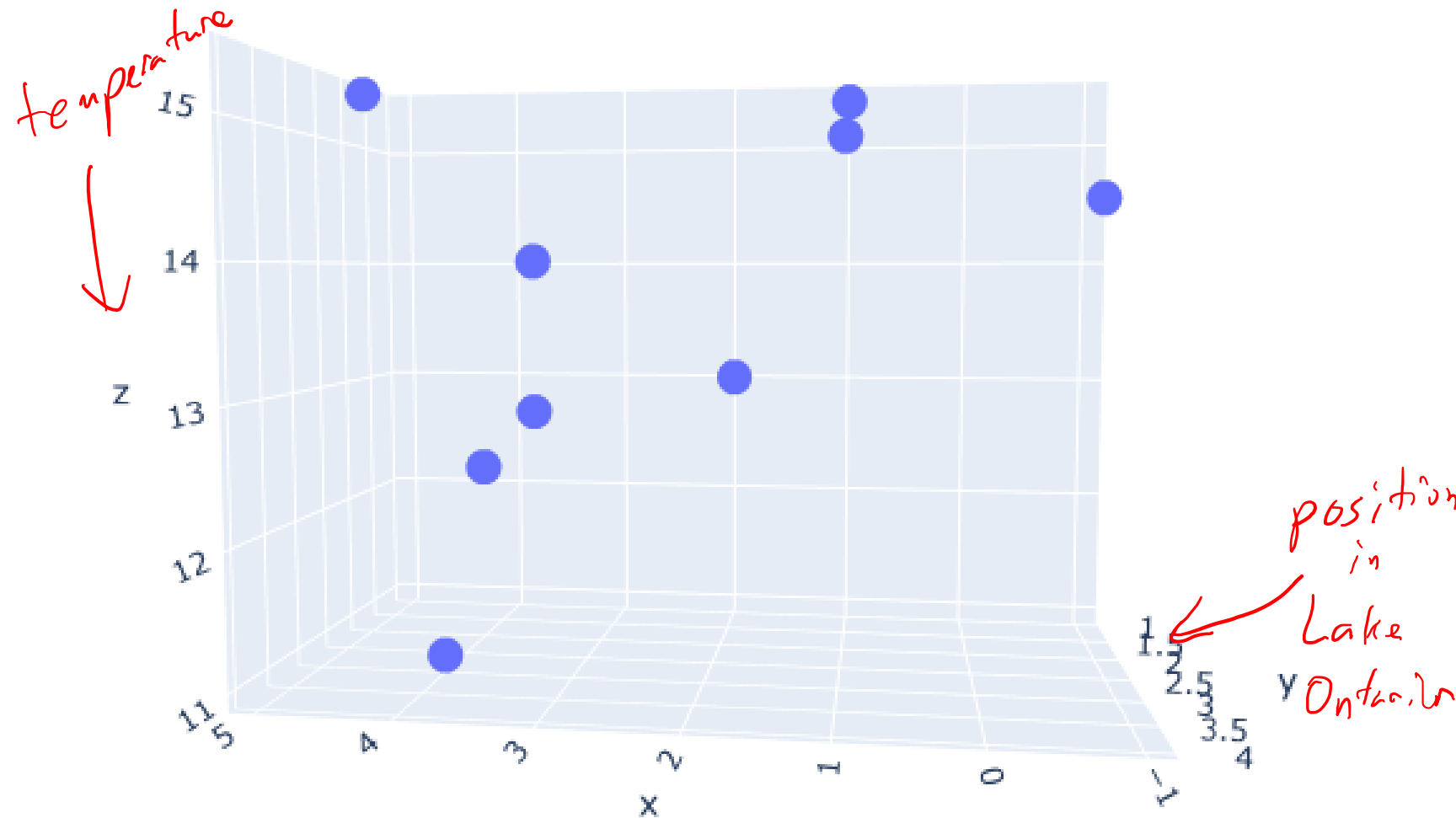


# Two-variable linear regression

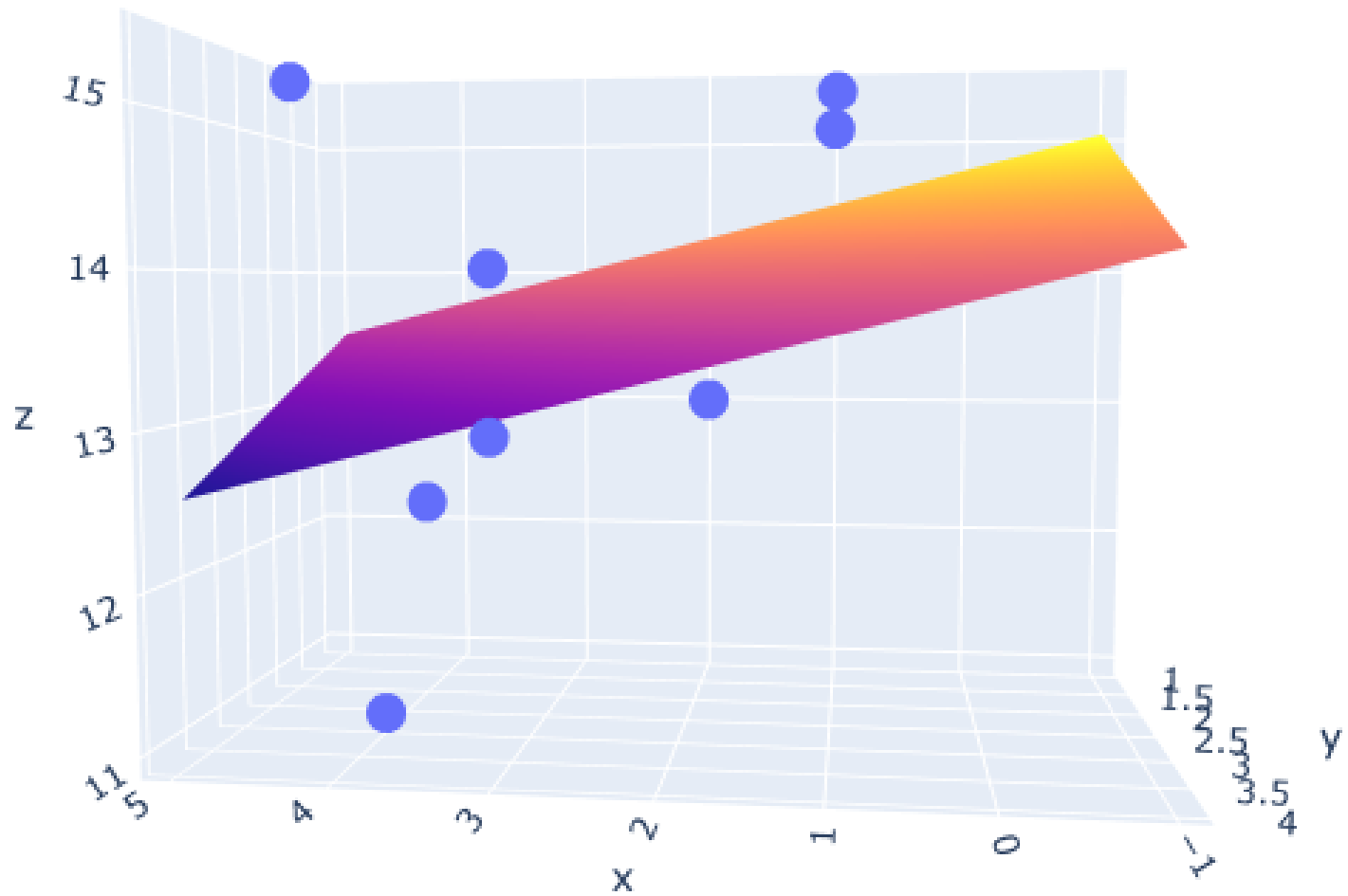
- What if we have multiple independent variables?
- Suppose we are measuring the water temperature in Lake Ontario, and want to know how the temperature varies as a function of location



# 3D Scatter Plot of temperatures



# Best-fit plane

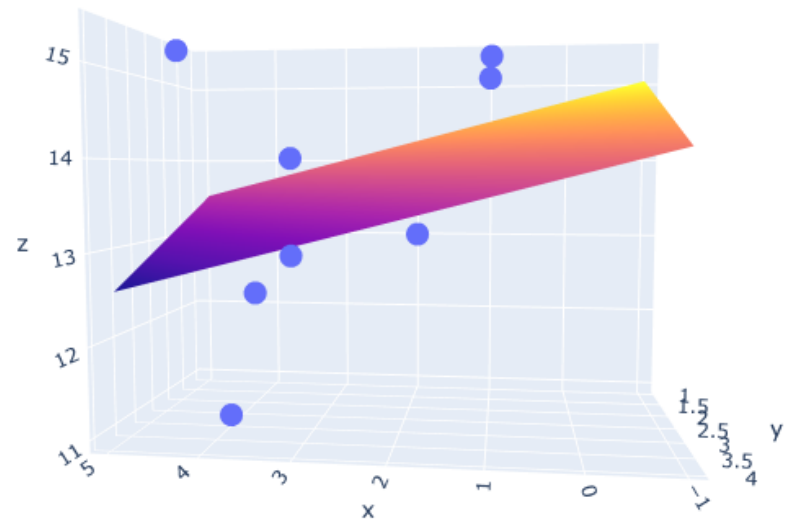
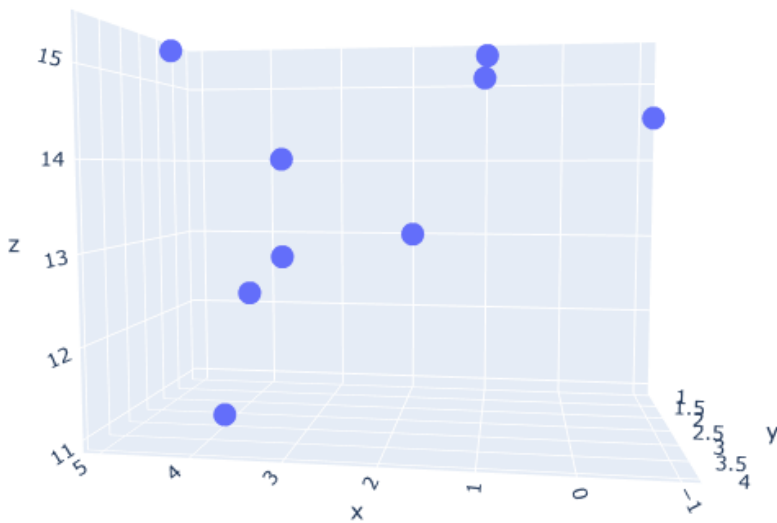


# Two-variable linear regression

- 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:*  $f(x, y) = m_1x + m_2y + b$   $f: \mathbb{R}^2 \rightarrow \mathbb{R}$

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

- Many independent variables, one dependent variable

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

- 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:

<sup>x</sup> Longitude	<sup>y</sup> Latitude	<sup>z</sup> 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

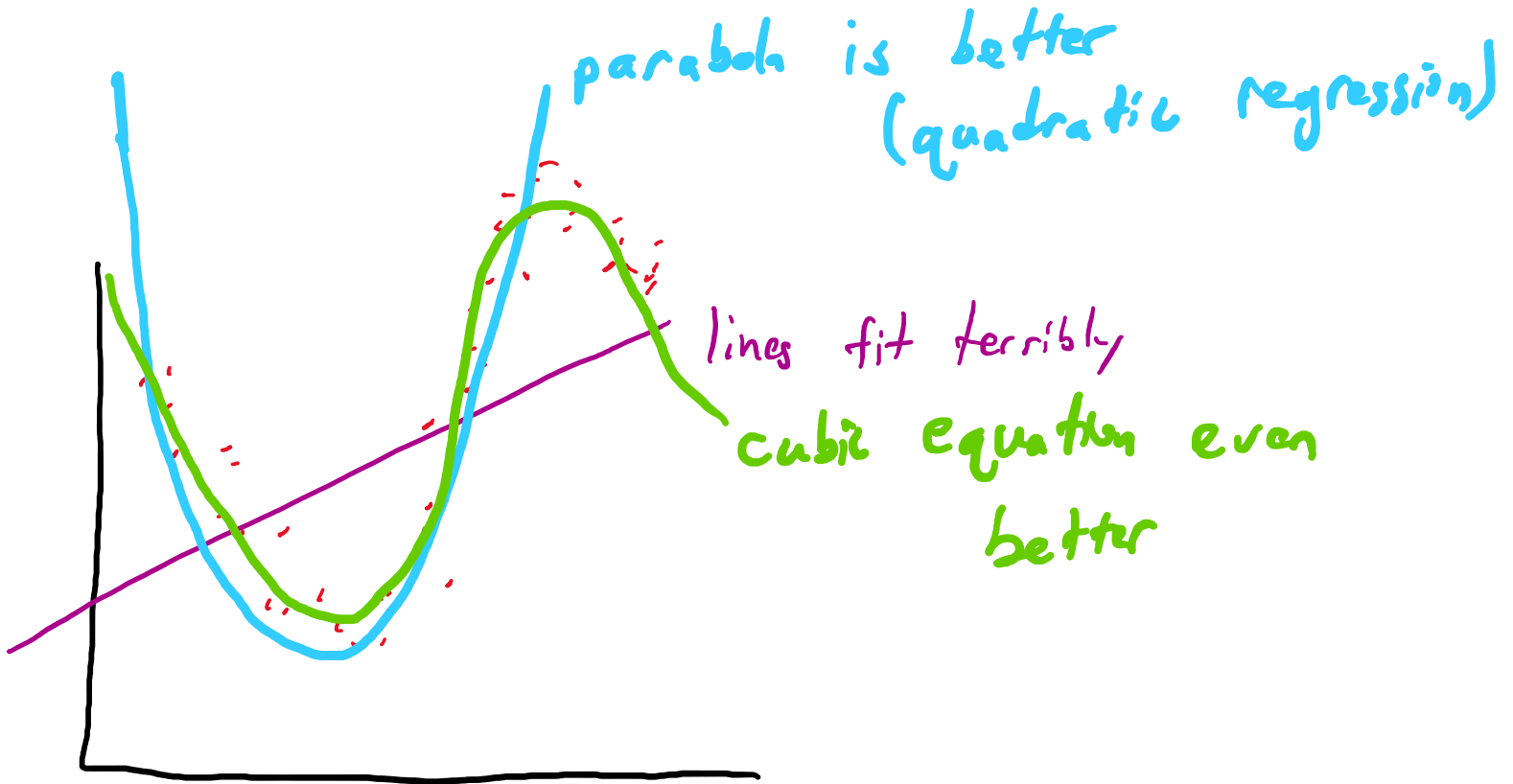
E: None of the  
above

- 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?



# Polynomial regression

- What if our data doesn't look linear?



# Different types of regression

- Linear regression:

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

- Quadratic regression:

$$f(x) = m_2x^2 + m_1x + b$$

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

- 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:

$$f(x) = b + m_0 e^{m_1 x}$$

# 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	$y = x^2$	$f(x)$
1	1	0.1
2	4	5.2
-1	1	6.9
4	16	4.2
⋮	⋮	⋮

$y = x^3$

Cubic:

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

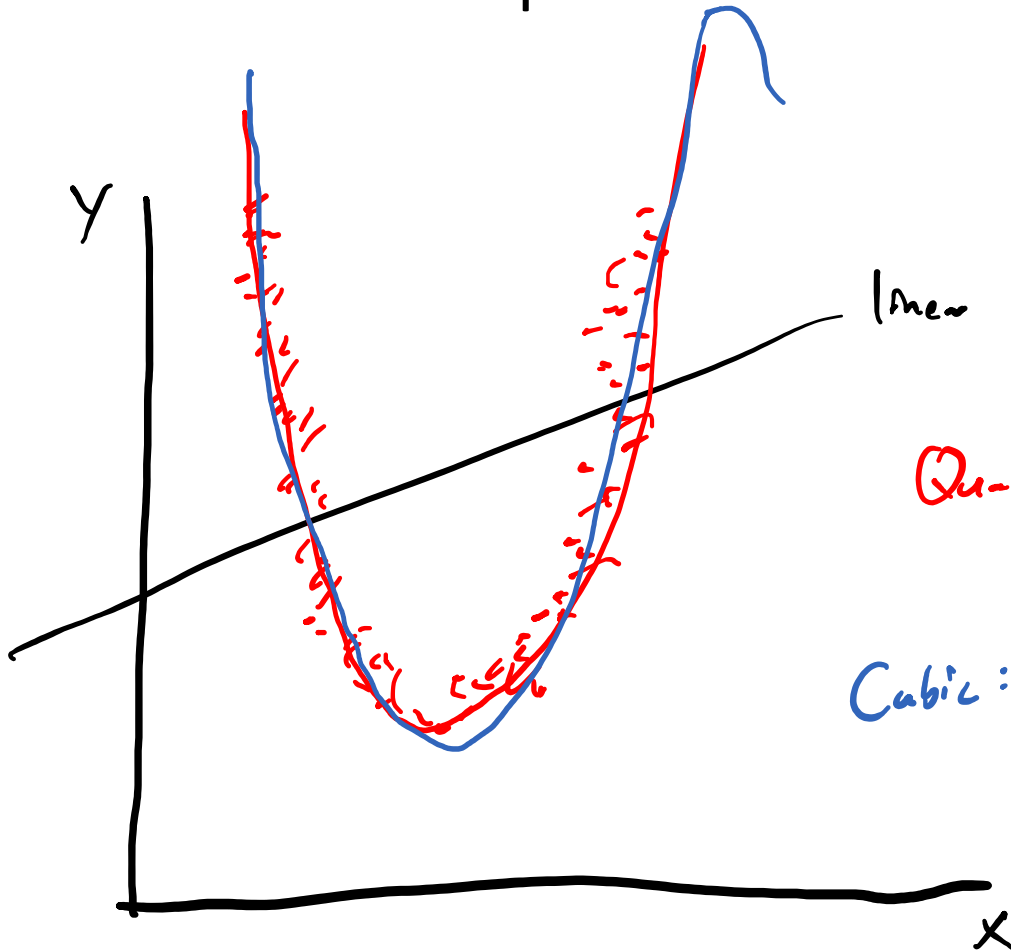
Let  $x_1 = x$     $x_2 = x^2$     $x_3 = x^3$

$$= m_1 x_1 + m_2 x_2 + m_3 x_3 + b$$

$$f\left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}\right) = \begin{bmatrix} m_1 & m_2 & m_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + b$$

# Intuition guess

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



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

Quadratic:  $f(x) = m_2x^2 + m_1x + b$

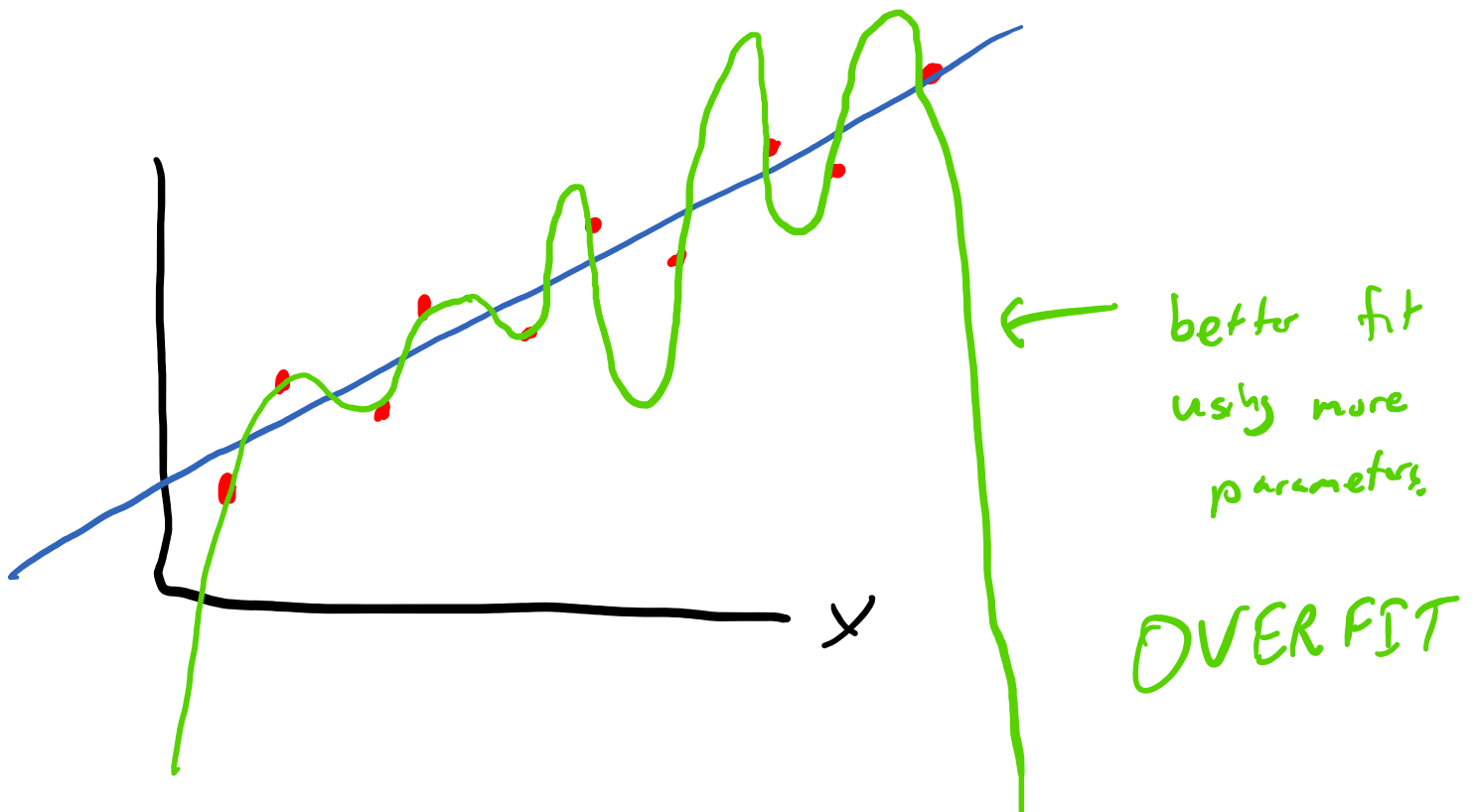
Cubic:  $f(x) = m_3x^3 + m_2x^2 + m_1x + b$

What if  $m_3 = 0$ ?

For same data, cubic will always have no higher error than quadratic for "training".

# 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. *on given data*
- 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”.



# Exponential regression

Exponential:

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

$$\ln f(x) = c_2 x + \ln c_1$$

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

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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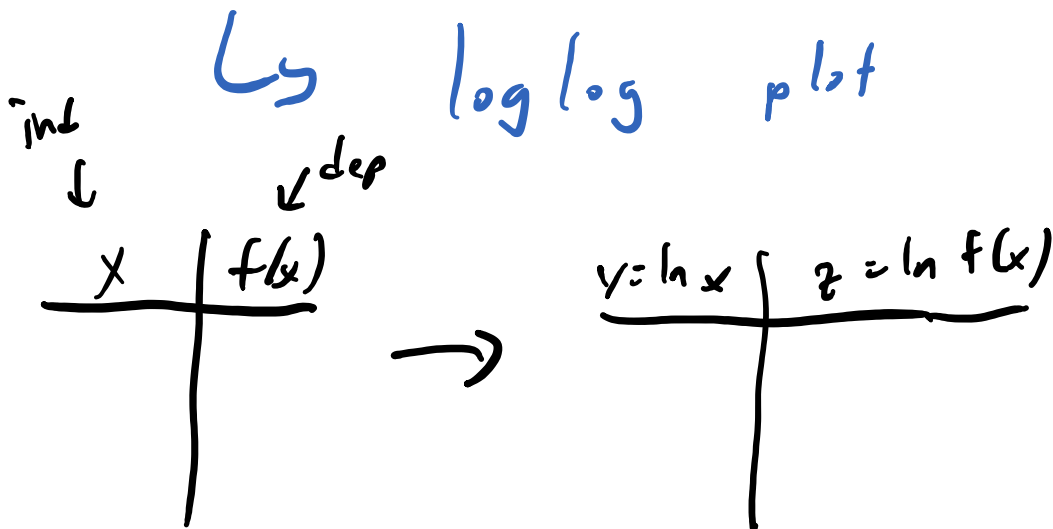
# Power dependencies

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

$$\ln f(x) = c_2 \ln x + \ln c_1$$

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

$$z = my + b$$



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# Power dependencies