

Introduction to the Mathematics of Regression Part 1

by

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for

Introduction to Artificial Intelligence
(CAP 4601)

Agenda

- Sum of Parabolas
- Minima: Weighted Sample Mean
- Sample Mean
- Sum of Squared Differences, Biased Sample Variance, and Unbiased Sample Variance
- Biased and Unbiased Sample Standard Deviation
- Sum of 2D Parabolas
- 2D Sample Mean
- Sum of the Product of Differences, Biased Sample Covariance, and Unbiased Sample Covariance
- Line
- Linear Regression
- Minima: Sum of Squared Differences and Sum of the Product of Differences
- Independent Variable Bias
- Online Mean
- One Way To Program Simple Linear Regression

Data

- Given:

{

0.255193,

2.23273,

3.85555,

4.49383,

4.90807,

6.24521,

6.41867,

8.09172,

8.61902,

9.39342

}



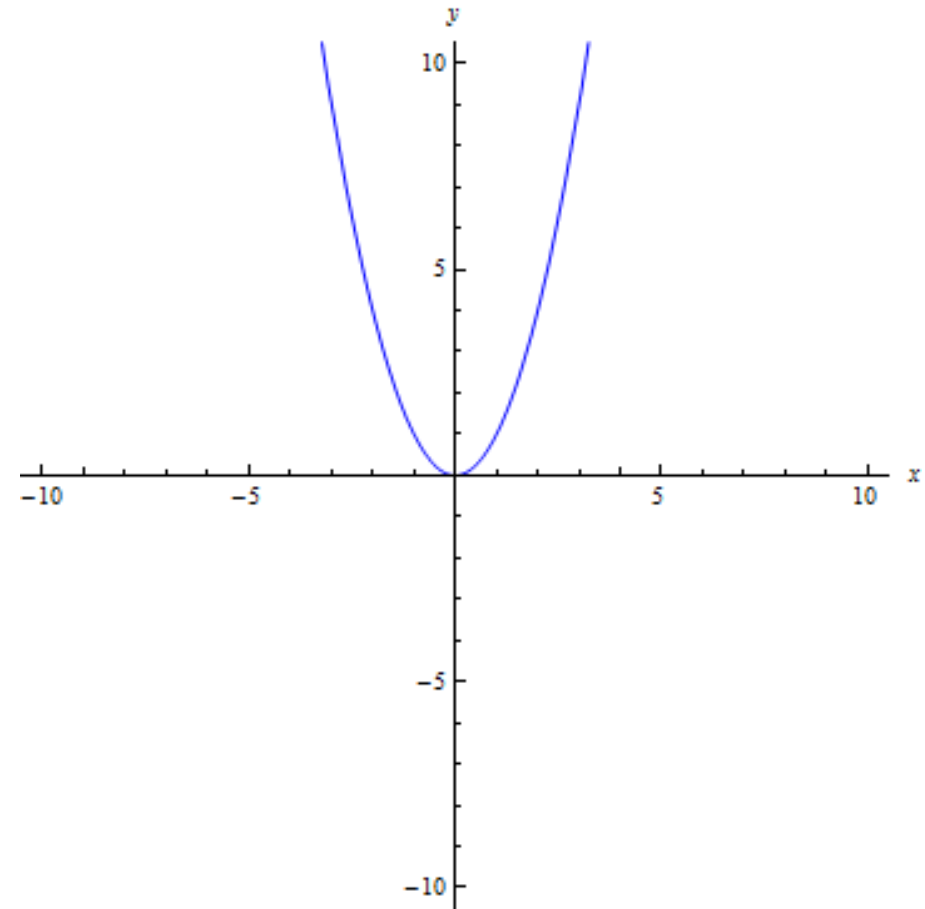
Parabola

$$y = w(x - x_0)^2 + y_0$$

Let $w = 1$, $x_0 = 0$, and $y_0 = 0$.

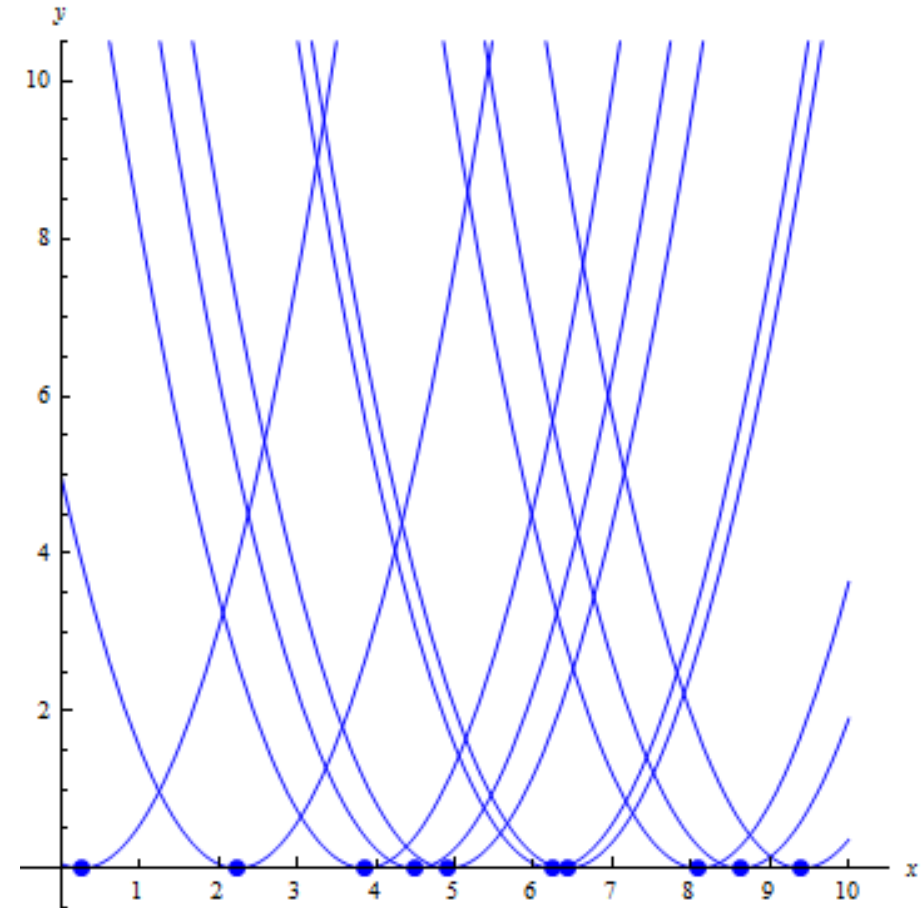
$$y = 1 \cdot (x - 0)^2 + 0$$

$$y = x^2$$



Parabolas

- For each datum, place a parabola centered on that value.



Sum of Parabolas

$$y = w(x - x_0)^2 + y_0 \quad \text{and} \quad y_0 = 0$$

$$y = \sum_{i=1}^n w_i (x - x_i)^2$$

Sum of Parabolas

$$w_1(x-x_1)^2 + w_2(x-x_2)^2 + w_3(x-x_3)^2 + w_4(x-x_4)^2 + w_5(x-x_5)^2 + \\ w_6(x-x_6)^2 + w_7(x-x_7)^2 + w_8(x-x_8)^2 + w_9(x-x_9)^2 + w_{10}(x-x_{10})^2$$

$$w_1x^2 - 2w_1x_1x + w_1x_1^2 + w_2x^2 - 2w_2x_2x + w_2x_2^2 + w_3x^2 - 2w_3x_3x + w_3x_3^2 + w_4x^2 - 2w_4x_4x + w_4x_4^2 + \\ w_5x^2 - 2w_5x_5x + w_5x_5^2 + w_6x^2 - 2w_6x_6x + w_6x_6^2 + w_7x^2 - 2w_7x_7x + w_7x_7^2 + w_8x^2 - 2w_8x_8x + w_8x_8^2 + \\ w_9x^2 - 2w_9x_9x + w_9x_9^2 + w_{10}x^2 - 2w_{10}x_{10}x + w_{10}x_{10}^2$$

$$(w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10})x^2 + \\ (-2w_1x_1 - 2w_2x_2 - 2w_3x_3 - 2w_4x_4 - 2w_5x_5 - 2w_6x_6 - 2w_7x_7 - 2w_8x_8 - 2w_9x_9 - 2w_{10}x_{10})x + \\ (w_1x_1^2 + w_2x_2^2 + w_3x_3^2 + w_4x_4^2 + w_5x_5^2 + w_6x_6^2 + w_7x_7^2 + w_8x_8^2 + w_9x_9^2 + w_{10}x_{10}^2)$$

Minima: Weighted Sample Mean

$$\frac{d}{dx} \left[\begin{array}{l} (w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10})x^2 + \\ (-2w_1x_1 - 2w_2x_2 - 2w_3x_3 - 2w_4x_4 - 2w_5x_5 - 2w_6x_6 - 2w_7x_7 - 2w_8x_8 - 2w_9x_9 - 2w_{10}x_{10})x + \\ (w_1x_1^2 + w_2x_2^2 + w_3x_3^2 + w_4x_4^2 + w_5x_5^2 + w_6x_6^2 + w_7x_7^2 + w_8x_8^2 + w_9x_9^2 + w_{10}x_{10}^2) \end{array} \right] = 0$$

$$\begin{array}{l} 2(w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10})x + \\ (-2w_1x_1 - 2w_2x_2 - 2w_3x_3 - 2w_4x_4 - 2w_5x_5 - 2w_6x_6 - 2w_7x_7 - 2w_8x_8 - 2w_9x_9 - 2w_{10}x_{10}) \end{array} = 0$$

$$x = \frac{-(-2w_1x_1 - 2w_2x_2 - 2w_3x_3 - 2w_4x_4 - 2w_5x_5 - 2w_6x_6 - 2w_7x_7 - 2w_8x_8 - 2w_9x_9 - 2w_{10}x_{10})}{2(w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10})}$$

$$x = \frac{w_1x_1 + w_2x_2 + w_3x_3 + w_4x_4 + w_5x_5 + w_6x_6 + w_7x_7 + w_8x_8 + w_9x_9 + w_{10}x_{10}}{w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10}}$$

Minima: Weighted Sample Mean

$$\frac{d^2}{dx^2} \left[\begin{aligned} & (w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10})x^2 + \\ & (-2w_1x_1 - 2w_2x_2 - 2w_3x_3 - 2w_4x_4 - 2w_5x_5 - 2w_6x_6 - 2w_7x_7 - 2w_8x_8 - 2w_9x_9 - 2w_{10}x_{10})x + \\ & (w_1x_1^2 + w_2x_2^2 + w_3x_3^2 + w_4x_4^2 + w_5x_5^2 + w_6x_6^2 + w_7x_7^2 + w_8x_8^2 + w_9x_9^2 + w_{10}x_{10}^2) \\ & 2(w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10}) \end{aligned} \right]$$

Since $0 < w_i$ for $i = 1, 2, \dots, 10$

$$0 < 2(w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10})$$

Sample Mean

Let $w_i = 1$ for $i = 1, 2, \dots, 10$.

$$x = \frac{w_1x_1 + w_2x_2 + w_3x_3 + w_4x_4 + w_5x_5 + w_6x_6 + w_7x_7 + w_8x_8 + w_9x_9 + w_{10}x_{10}}{w_1 + w_2 + w_3 + w_4 + w_5 + w_6 + w_7 + w_8 + w_9 + w_{10}}$$

$$x = \frac{1x_1 + 1x_2 + 1x_3 + 1x_4 + 1x_5 + 1x_6 + 1x_7 + 1x_8 + 1x_9 + 1x_{10}}{1+1+1+1+1+1+1+1+1+1}$$

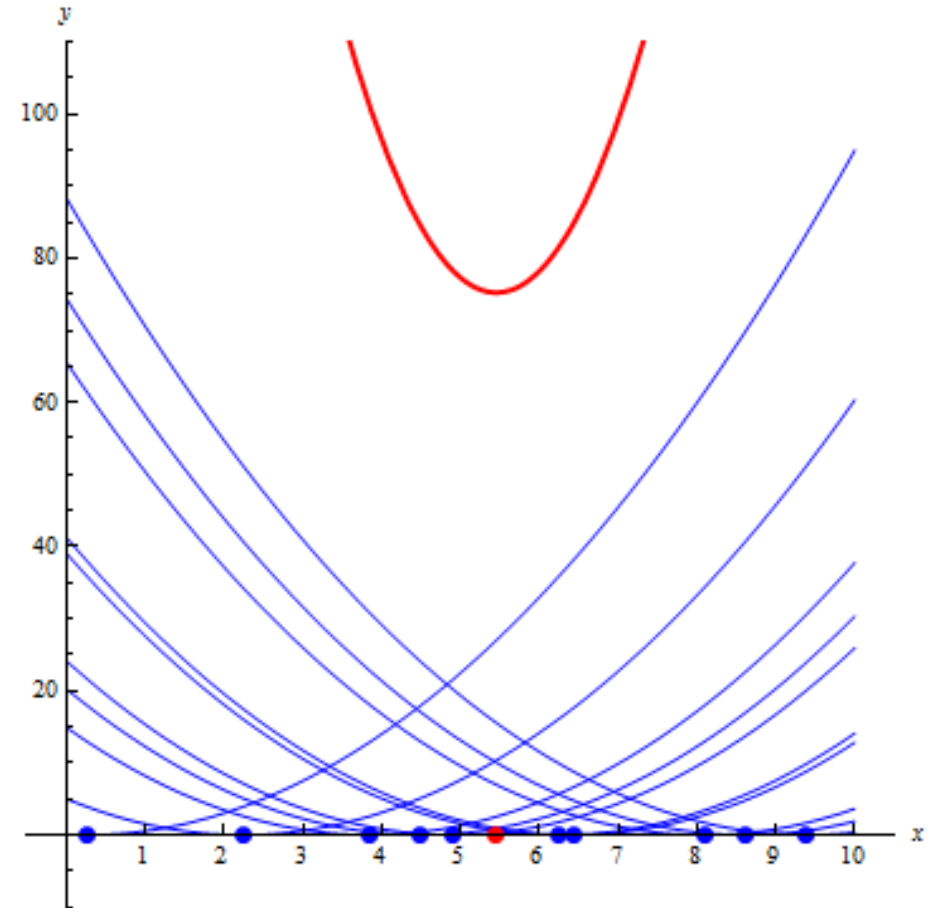
$$x = \frac{x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}}{10}$$

Sample Mean

$$\bar{x} = \frac{\sum_{i=1}^n x_i}{n}$$

Sample Mean

$$\bar{x} = 5.45134$$

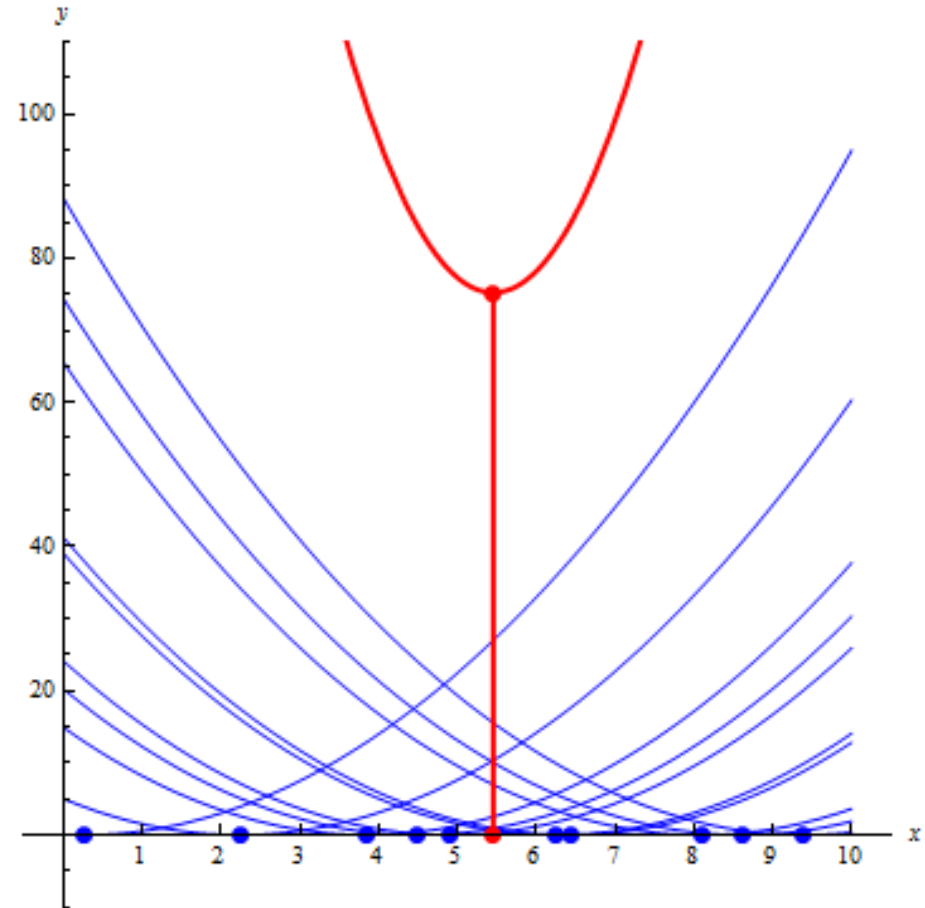


Sum of Parabolas: Sum of Squared Differences

$$y = \sum_{i=1}^n w_i (x - x_i)^2$$

Let $x = \bar{x}$ and $w_i = 1$, then

$$y = \sum_{i=1}^n (\bar{x} - x_i)^2$$



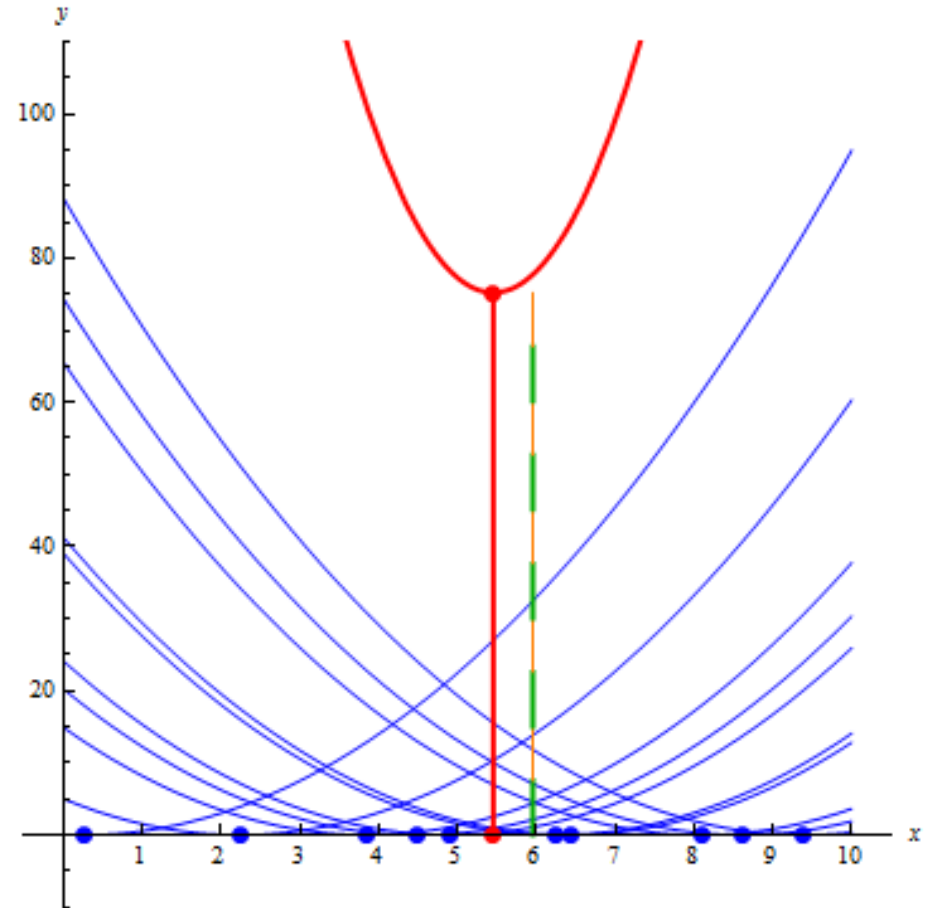
Sum of Parabolas: Biased Sample Variance

$$y = \sum_{i=1}^n w_i (x - x_i)^2$$

Let $x = \bar{x}$ and $w_i = \frac{1}{n}$, then

$$y = \frac{1}{n} \sum_{i=1}^n (\bar{x} - x_i)^2$$

$$s_{\text{biased}}^2 = 7.52297$$



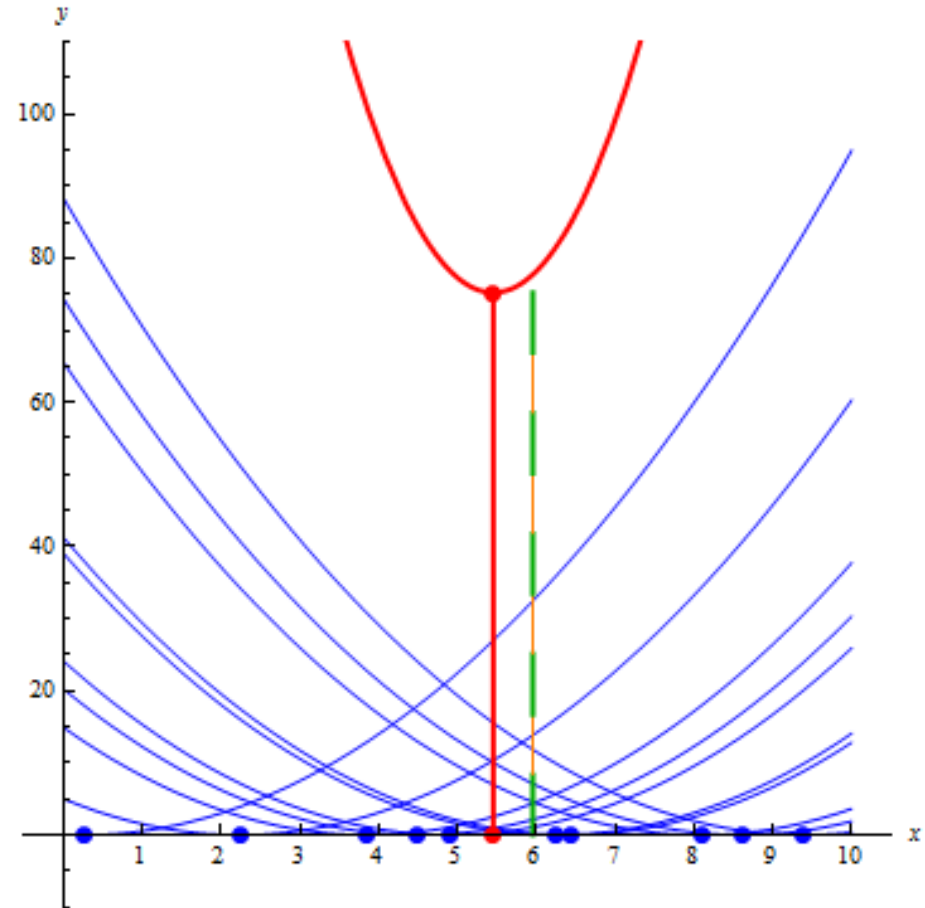
Sum of Parabolas: Unbiased Sample Variance

$$y = \sum_{i=1}^n w_i (x - x_i)^2$$

Let $x = \bar{x}$ and $w_i = \frac{1}{n-1}$, then

$$y = \frac{1}{n-1} \sum_{i=1}^n (\bar{x} - x_i)^2$$

$$s_{\text{unbiased}}^2 = 8.358885$$



Biased and Unbiased Sample Variance

$$y = \sum_{i=1}^n w_i (x - x_i)^2$$

If $x = \bar{x}$ and $w_i = \frac{1}{n}$, then

$$s^2_{\text{biased}} = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2$$

If $x = \bar{x}$ and $w_i = \frac{1}{n-1}$, then

$$s^2_{\text{unbiased}} = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

Sample Standard Deviation

- Biased:

$$\sqrt{s^2_{\text{biased}}} = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2}$$

$$s_{\text{biased}} = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2}$$

- Unbiased:

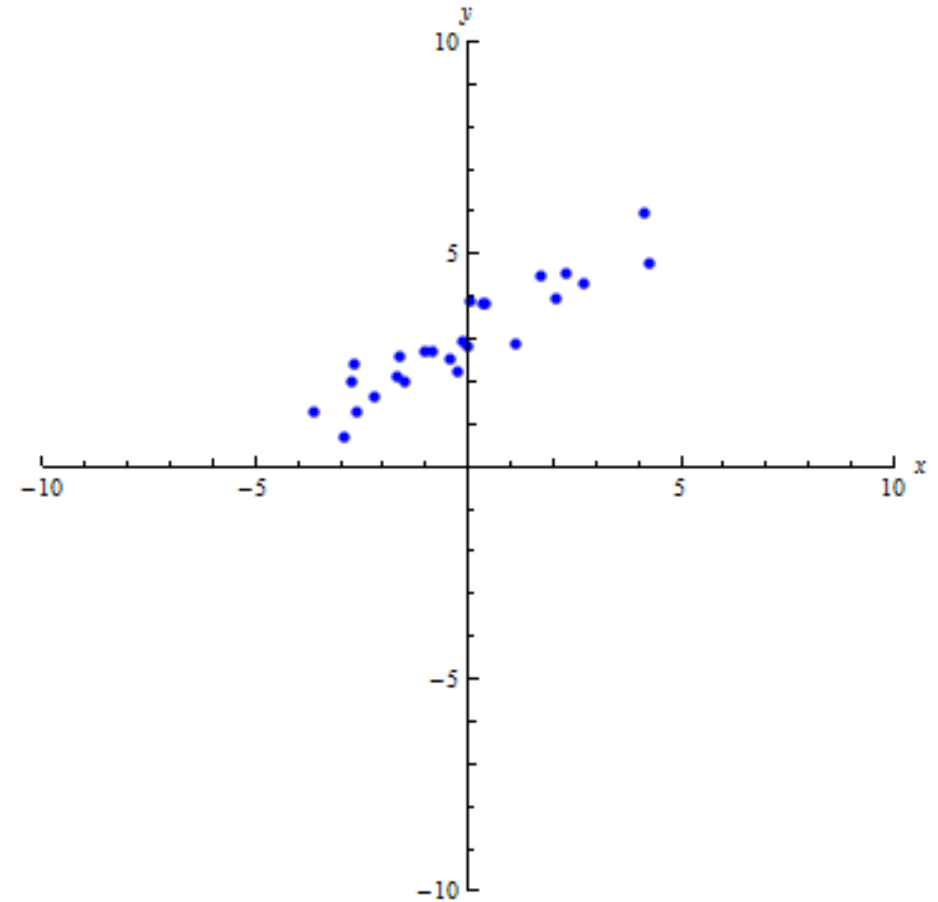
$$\sqrt{s^2_{\text{unbiased}}} = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$$

$$s_{\text{unbiased}} = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$$

Data

- Data:

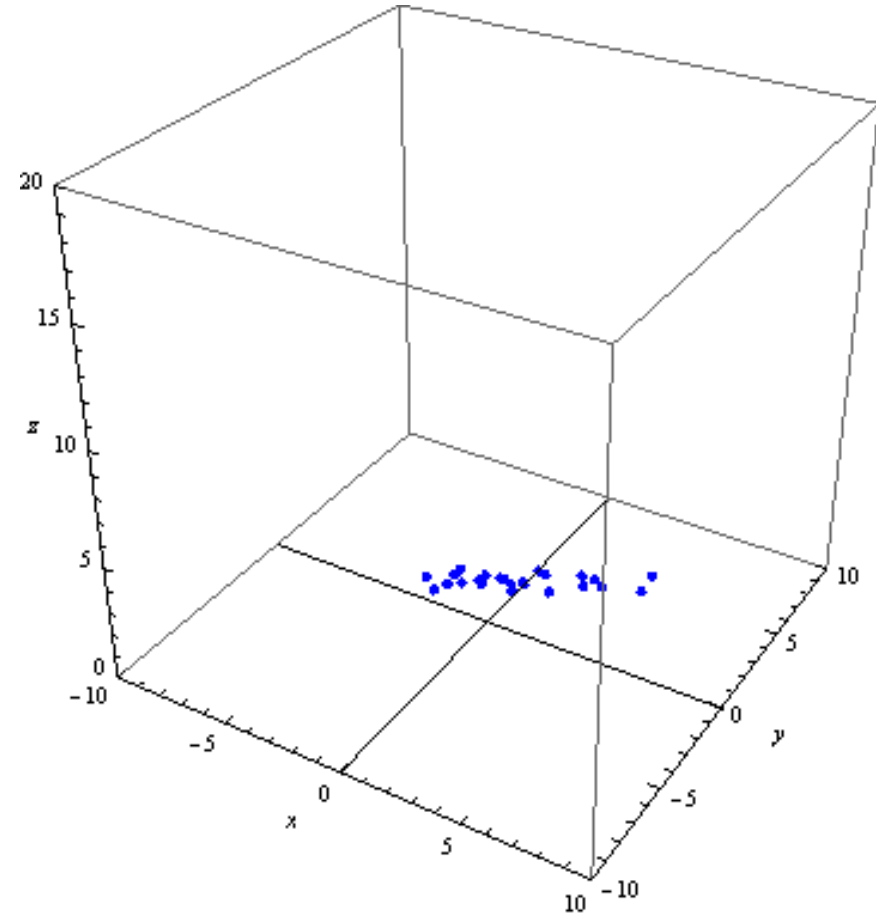
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  {1.10033, 2.89639},  
  {-1.63906, 2.56916},  
  {2.27983, 4.57127},  
  {-0.836348, 2.70824},  
  {-2.90988, 0.685828},  
  {-0.104817, 2.95222},  
  {-0.226538, 2.24849},  
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  {-0.00953108, 2.84022},  
  {0.336282, 3.86626},  
  {4.12633, 5.94993},  
  {1.70053, 4.51368},  
  {-1.4793, 1.99986},  
  {0.0467884, 3.91811},  
  {-1.70285, 2.12758},  
  {-3.61035, 1.26436},  
  {2.08504, 3.94459},  
  {4.23512, 4.80965},  
  {2.70993, 4.30984},  
  {-2.72741, 1.97363},  
  {-0.418723, 2.55797}  
}
```



Data

- Data:

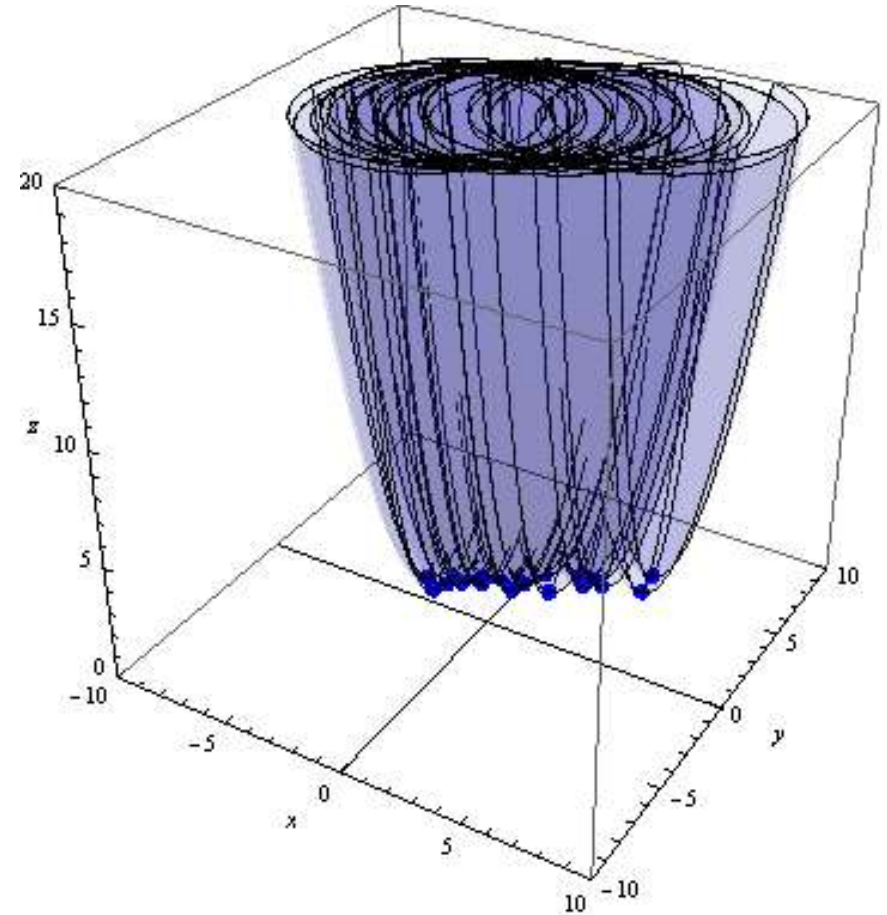
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  {1.10033, 2.89639},  
  {-1.63906, 2.56916},  
  {2.27983, 4.57127},  
  {-0.836348, 2.70824},  
  {-2.90988, 0.685828},  
  {-0.104817, 2.95222},  
  {-0.226538, 2.24849},  
  {-2.64364, 1.28981},  
  {-0.00953108, 2.84022},  
  {0.336282, 3.86626},  
  {4.12633, 5.94993},  
  {1.70053, 4.51368},  
  {-1.4793, 1.99986},  
  {0.0467884, 3.91811},  
  {-1.70285, 2.12758},  
  {-3.61035, 1.26436},  
  {2.08504, 3.94459},  
  {4.23512, 4.80965},  
  {2.70993, 4.30984},  
  {-2.72741, 1.97363},  
  {-0.418723, 2.55797}  
}
```



2D Parabolas

- 2D Parabola:

$$z = (x - x_0)^2 + (y - y_0)^2$$



Sum of 2D Parabolas

$$f(x, y) = \sum_{i=1}^n \left((x - x_i)^2 + (y - y_i)^2 \right) = \sum_{i=1}^n (x - x_i)^2 + \sum_{i=1}^n (y - y_i)^2$$

$$\frac{\partial f}{\partial x} = \sum_{i=1}^n 2(x - x_i) \quad \frac{\partial^2 f}{\partial x^2} = \sum_{i=1}^n 2 = 2n \quad \frac{\partial^2 f}{\partial x \partial y} = 0$$

$$\frac{\partial f}{\partial y} = \sum_{i=1}^n 2(y - y_i) \quad \frac{\partial^2 f}{\partial y \partial x} = 0 \quad \frac{\partial^2 f}{\partial y^2} = \sum_{i=1}^n 2 = 2n$$

Extrema: First Derivative

$$\frac{\partial f}{\partial x} = 0 \quad \text{and} \quad \frac{\partial f}{\partial x} = \sum_{i=1}^n 2(x - x_i)$$

$$\sum_{i=1}^n 2(x - x_i) = 0 \quad \sum_{i=1}^n x - \sum_{i=1}^n x_i = 0 \quad nx = \sum_{i=1}^n x_i$$

$$\sum_{i=1}^n (x - x_i) = 0 \quad \sum_{i=1}^n x = \sum_{i=1}^n x_i \quad x = \frac{\sum_{i=1}^n x_i}{n}$$

Extrema: First Derivative

$$\frac{\partial f}{\partial y} = 0 \quad \text{and} \quad \frac{\partial f}{\partial y} = \sum_{i=1}^n 2(y - y_i)$$

$$\sum_{i=1}^n 2(y - y_i) = 0 \quad \sum_{i=1}^n y - \sum_{i=1}^n y_i = 0 \quad ny = \sum_{i=1}^n y_i$$

$$\sum_{i=1}^n (y - y_i) = 0 \quad \sum_{i=1}^n y = \sum_{i=1}^n y_i \quad y = \frac{\sum_{i=1}^n y_i}{n}$$

Extrema: Second Derivative (Hessian Matrix)

Minima if:

$$\begin{pmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{pmatrix} \quad 0 < \left| \left(\frac{\partial^2 f}{\partial x^2} \right) \right| \quad \text{and} \quad 0 < \left| \begin{pmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{pmatrix} \right|$$
$$0 < \frac{\partial^2 f}{\partial x^2} \quad \text{and} \quad 0 < \frac{\partial^2 f}{\partial x^2} \frac{\partial^2 f}{\partial y^2} - \frac{\partial^2 f}{\partial x \partial y} \frac{\partial^2 f}{\partial y \partial x}$$

Extrema: Second Derivative (Hessian Matrix)

Minima if:

$$\begin{pmatrix} 2n & 0 \\ 0 & 2n \end{pmatrix}$$
$$0 < |(2n)| \quad \text{and} \quad 0 < \begin{vmatrix} 2n & 0 \\ 0 & 2n \end{vmatrix}$$
$$0 < 2n \quad \text{and} \quad 0 < (2n)(2n) - (0)(0)$$
$$0 < 2n \quad \text{and} \quad 0 < (2n)^2$$

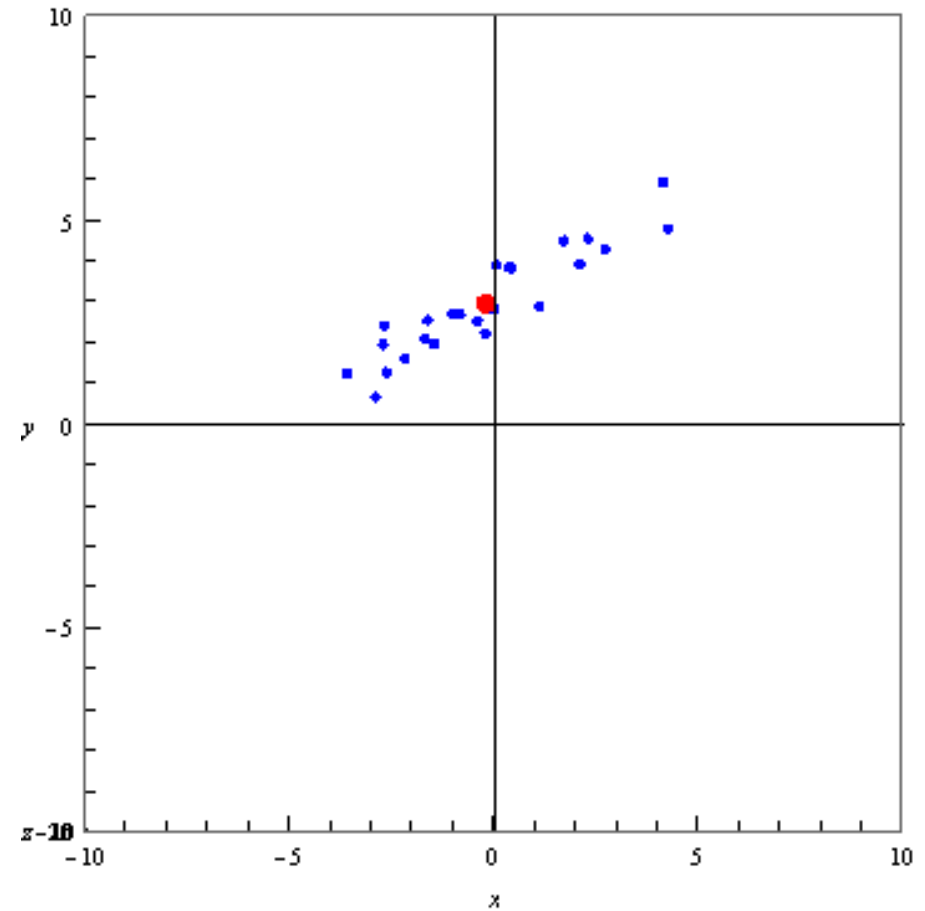
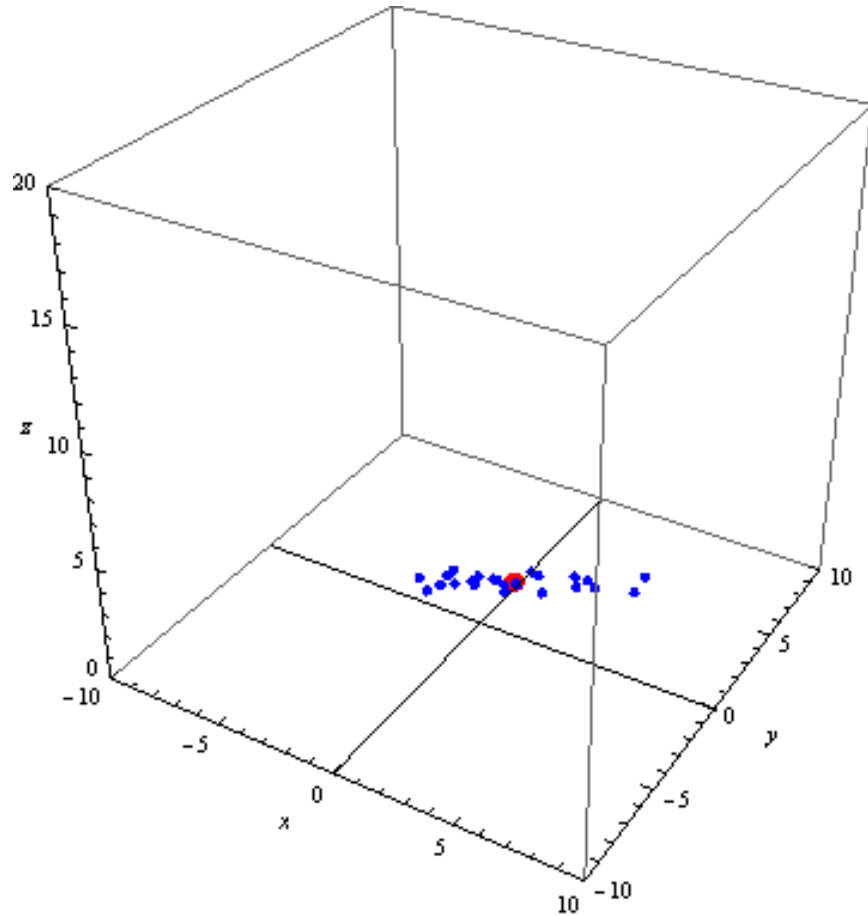
Minima: Sample Mean

Therefore, we have a minima at:

$$x = \frac{\sum_{i=1}^n x_i}{n} \qquad y = \frac{\sum_{i=1}^n y_i}{n}$$

$$\bar{x} = \frac{\sum_{i=1}^n x_i}{n} \qquad \bar{y} = \frac{\sum_{i=1}^n y_i}{n}$$

Sample Mean

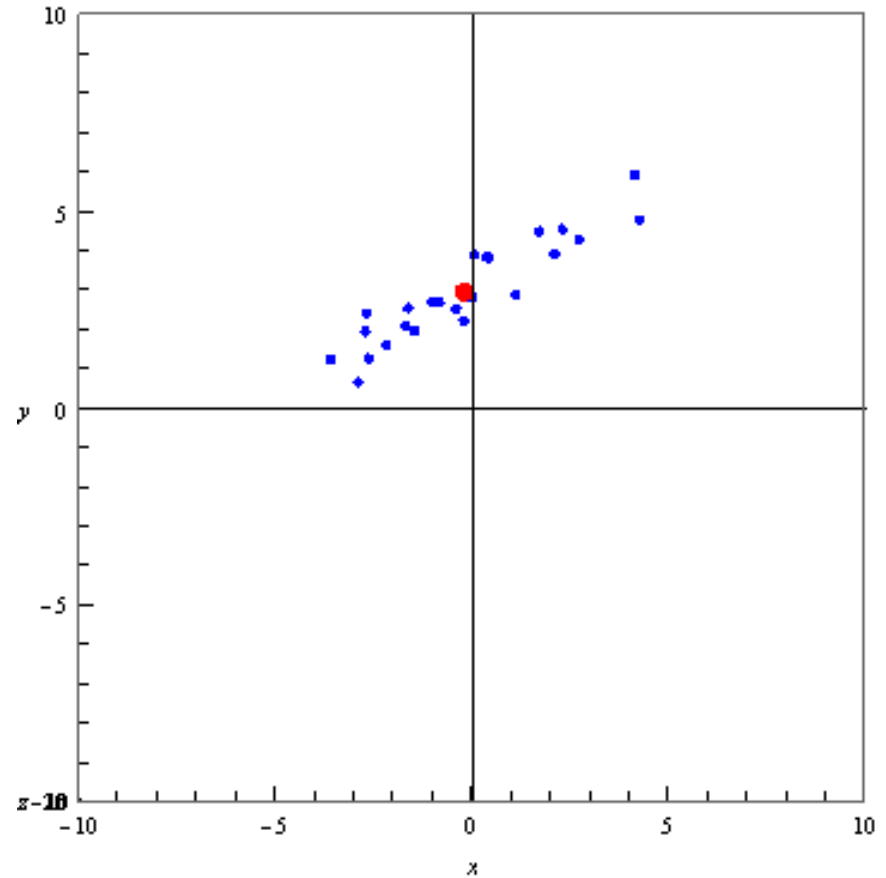


Sample Mean

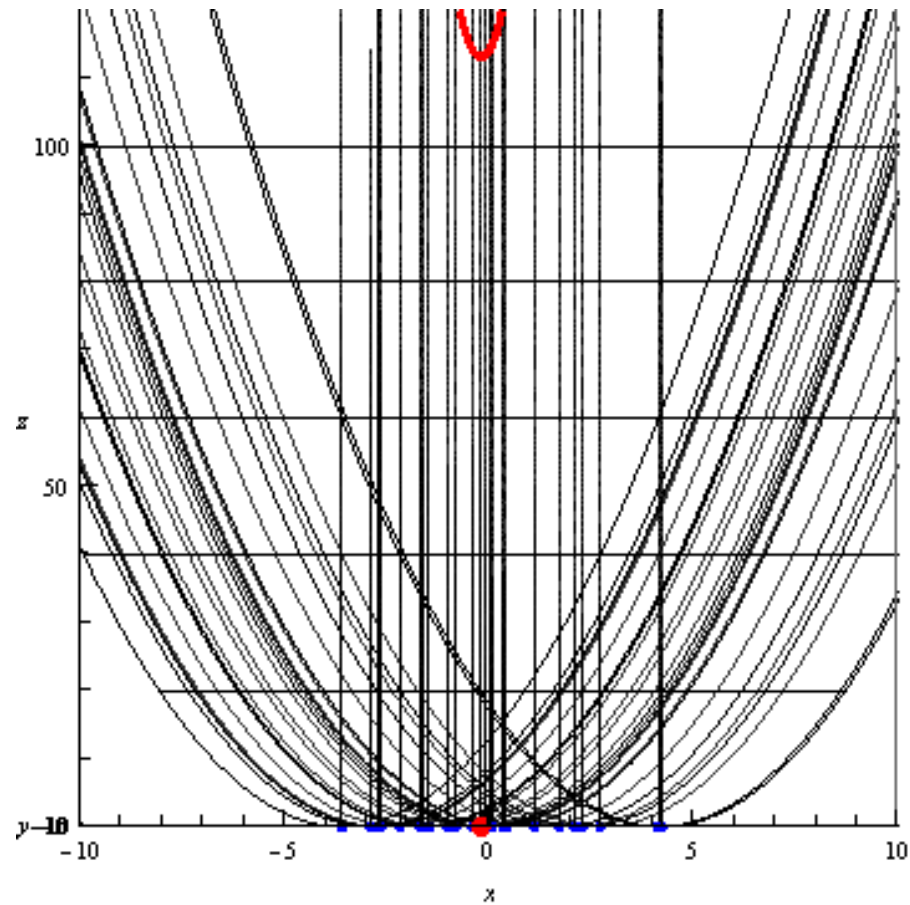
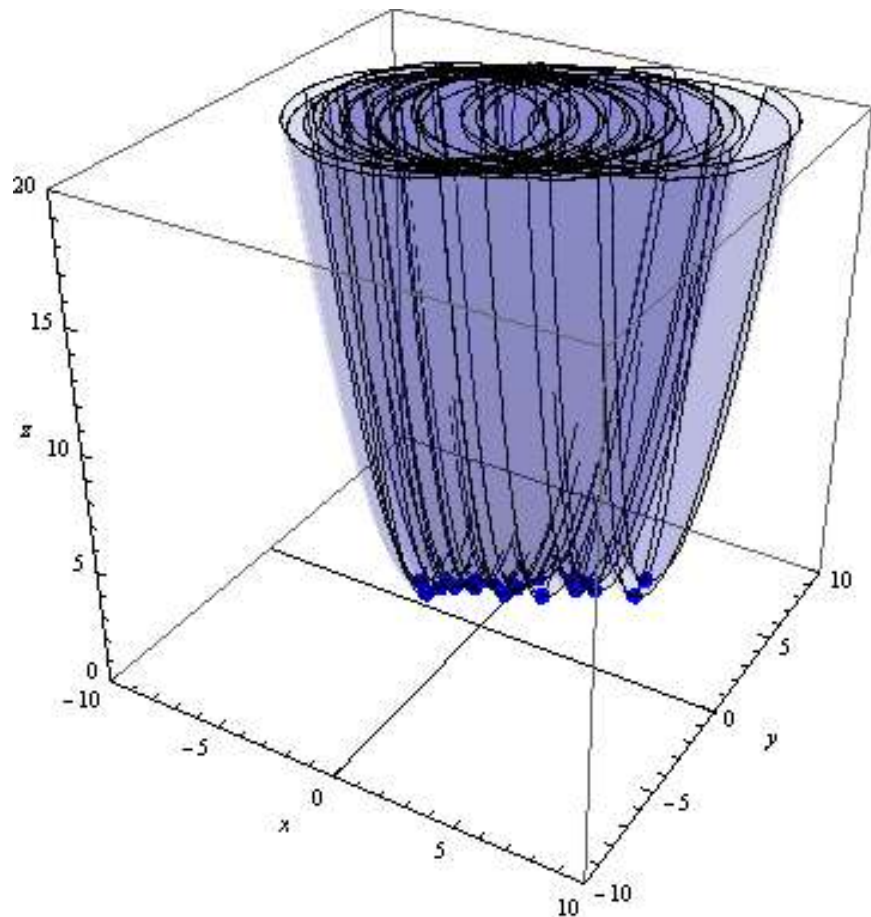
$$\bar{x} = -0.208094$$

$$\bar{y} = 2.98517$$

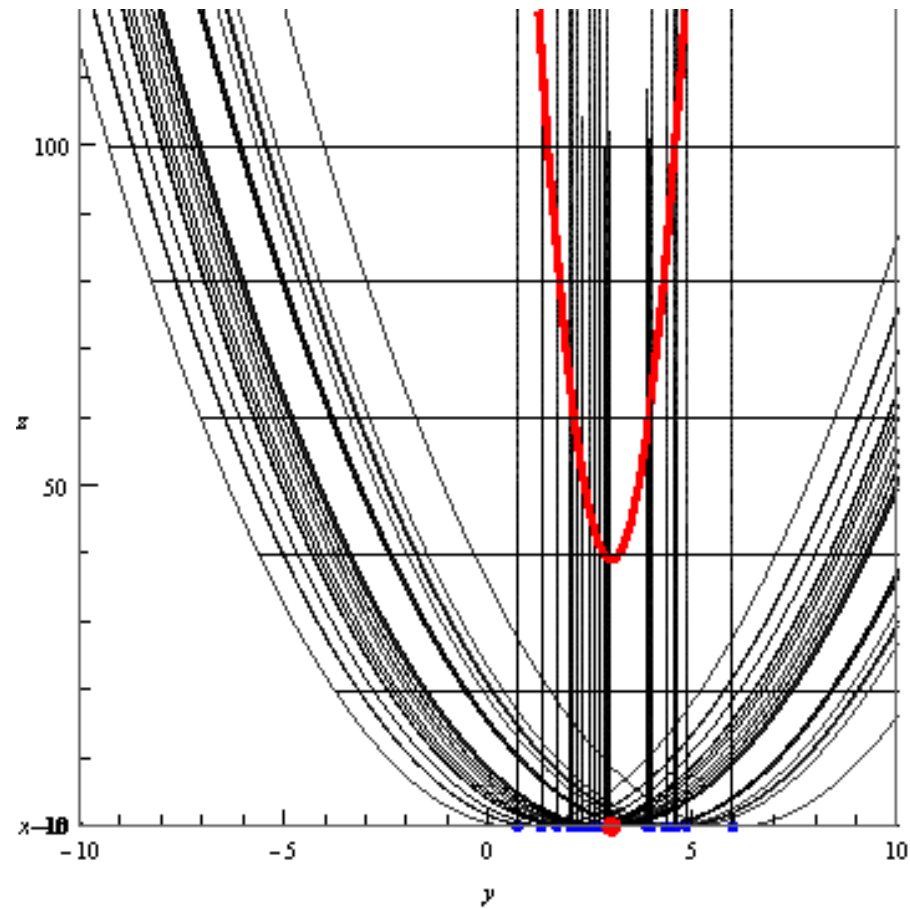
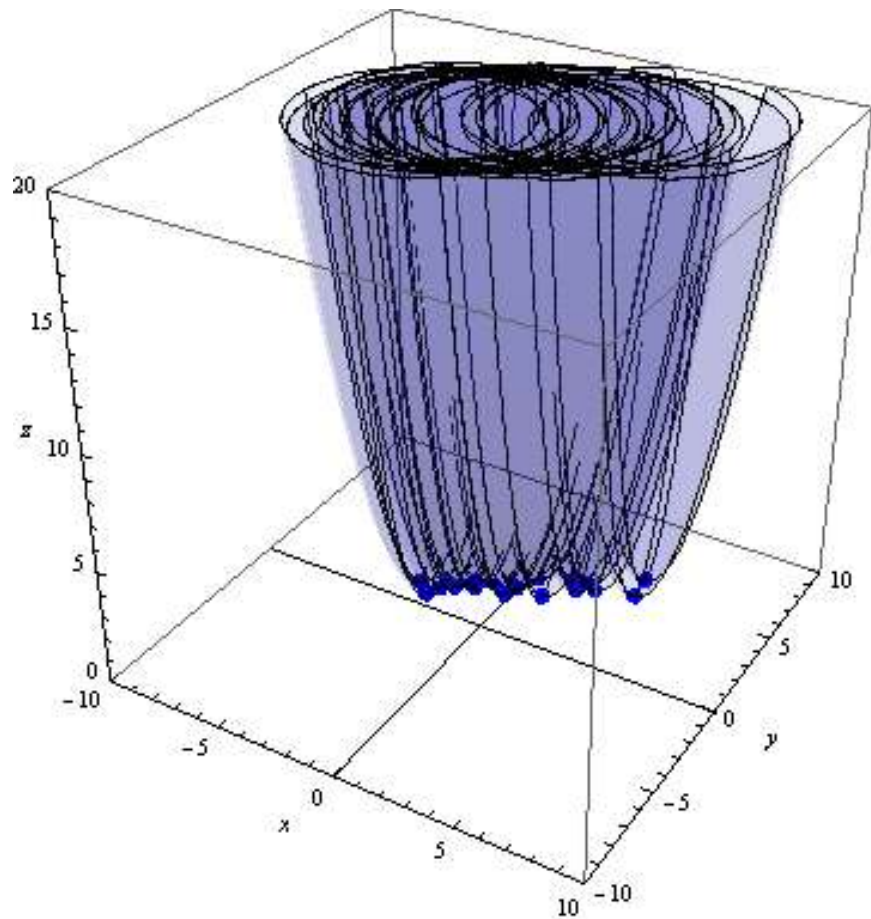
$$(-0.208094, 2.98517)$$



Sum of Parabolas: x



Sum of Parabolas: y



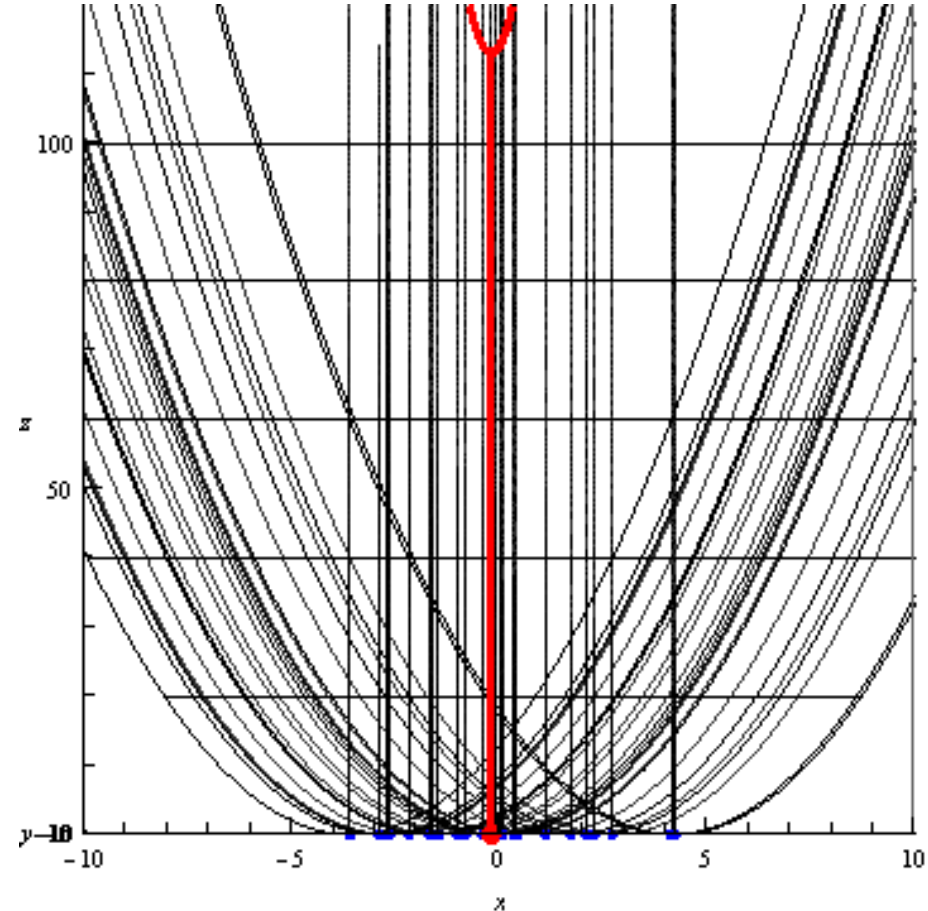
Sum of Squared Differences: x

$$z_{xx} = \sum_{i=1}^n (x - x_i)^2$$

Let $x = \bar{x}$, then

$$z_{xx} = \sum_{i=1}^n (\bar{x} - x_i)^2$$

$$z_{xx} = 112.937$$



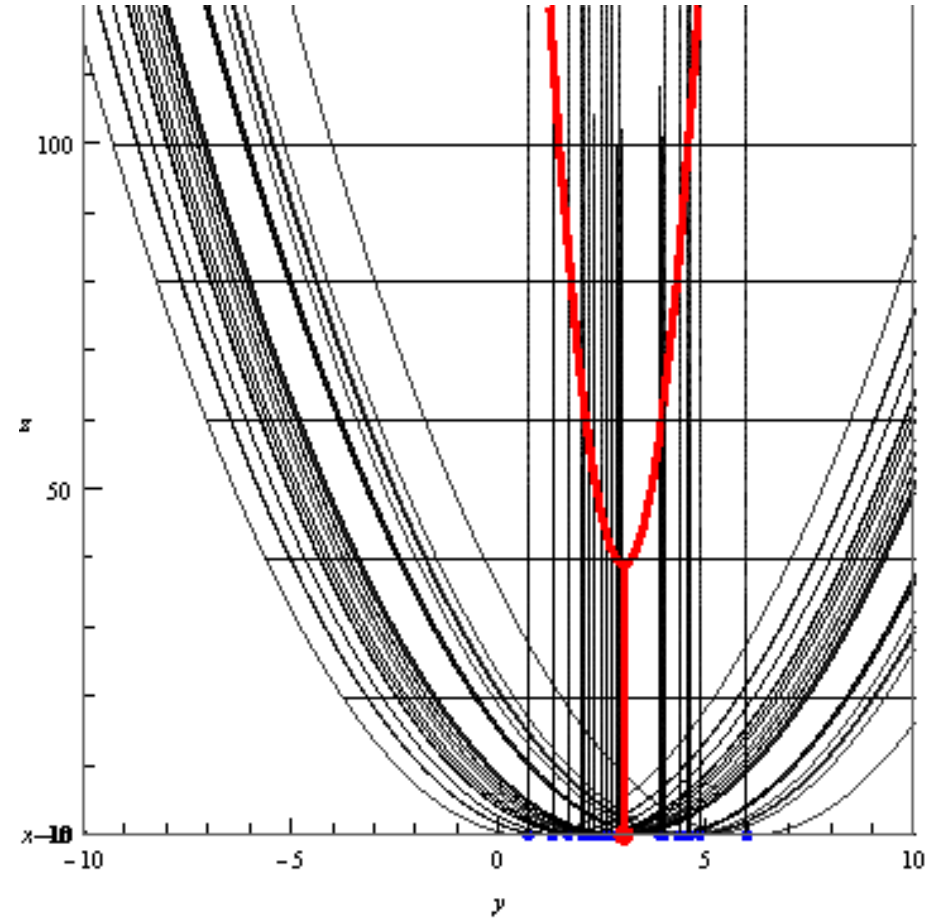
Sum of Squared Differences: y

$$z_{yy} = \sum_{i=1}^n (y - y_i)^2$$

Let $y = \bar{y}$, then

$$z_{yy} = \sum_{i=1}^n (\bar{y} - y_i)^2$$

$$z_{yy} = 39.096$$



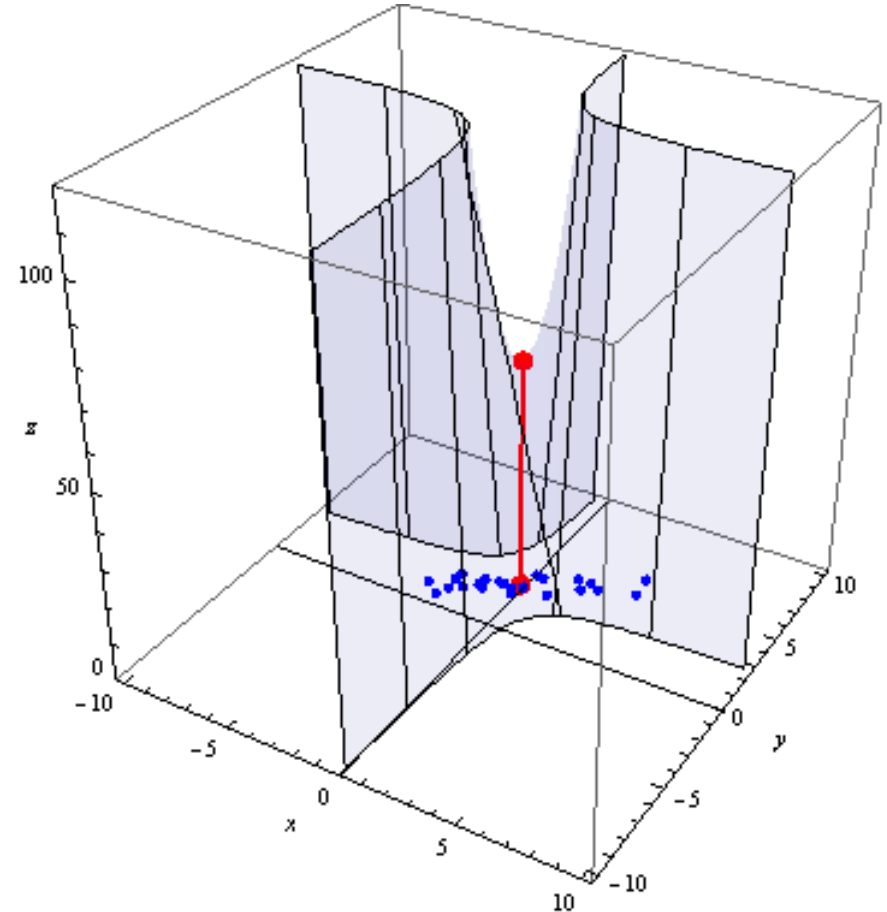
Sum of the Product of Differences

$$z_{xy} = \sum_{i=1}^n (x - x_i)(y - y_i)$$

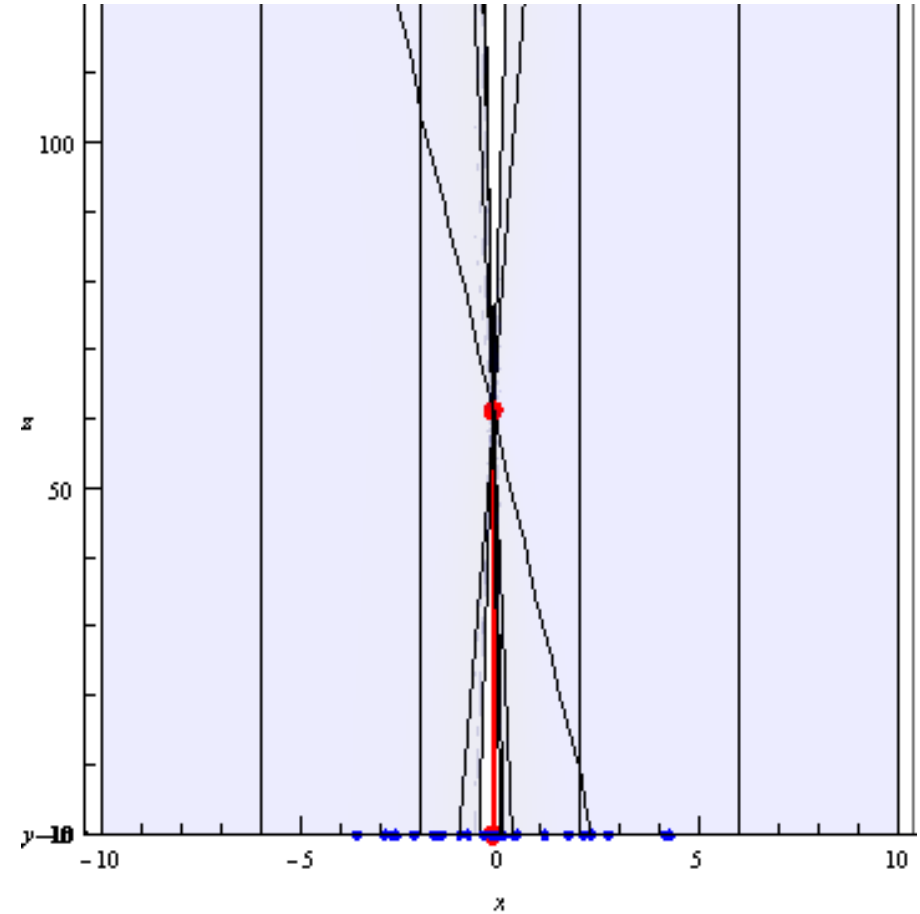
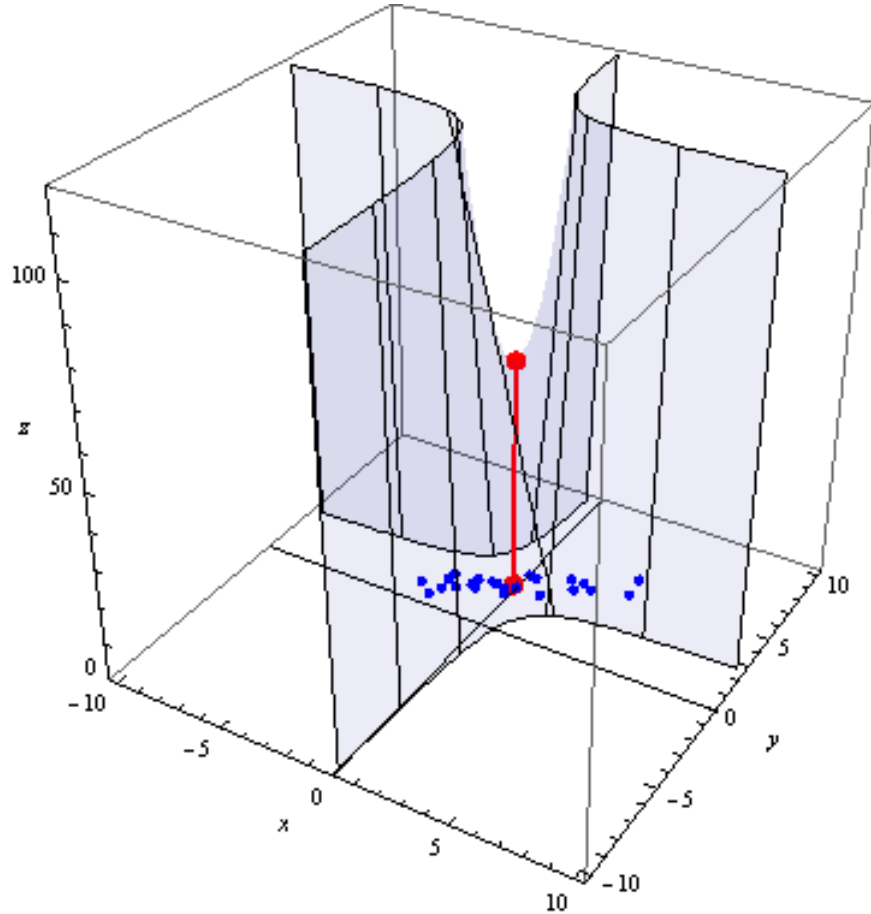
Let $x = \bar{x}$ and $y = \bar{y}$, then

$$z_{xy} = \sum_{i=1}^n (\bar{x} - x_i)(\bar{y} - y_i)$$

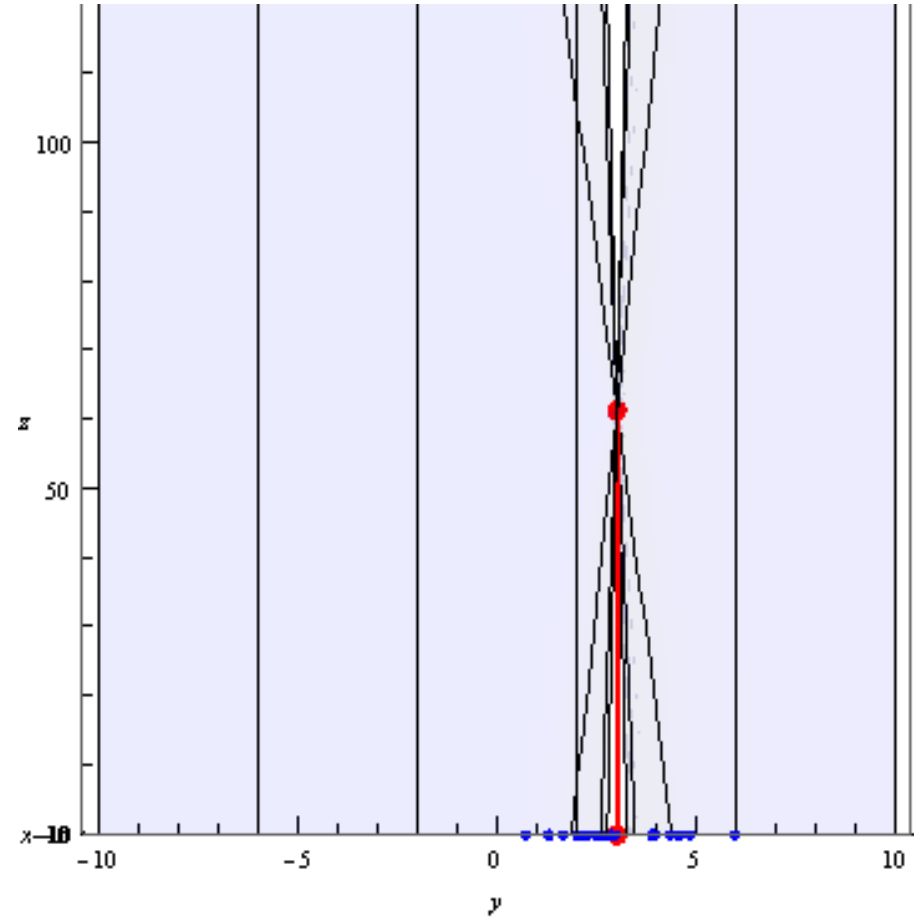
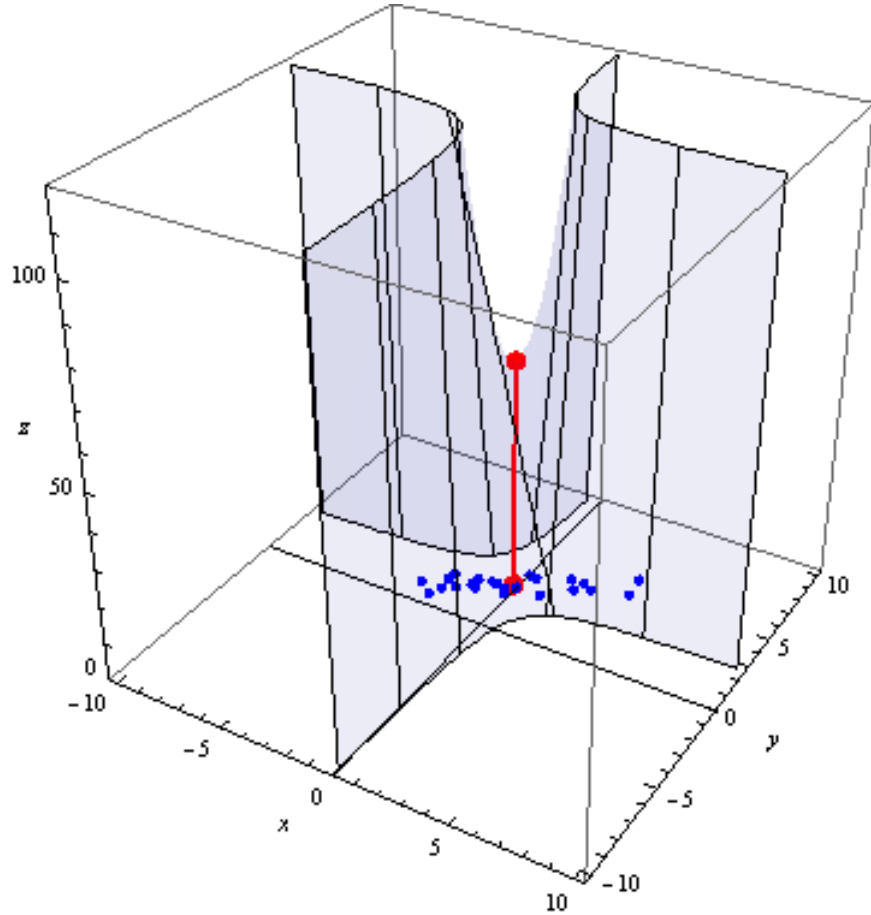
$$z_{xy} = 61.4087$$



Sum of the Product of Differences: x



Sum of the Product of Differences: y



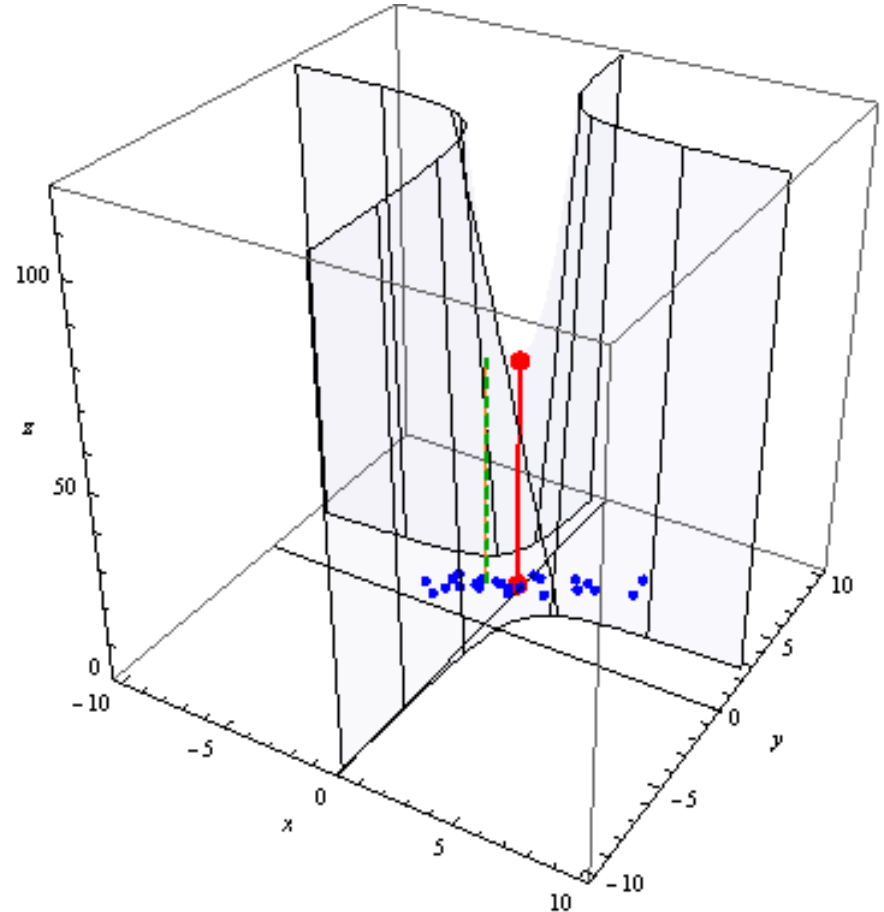
Sum of the Product of Differences: Biased Sample Covariance

$$z_{xy} = \sum_{i=1}^n w_i (x - x_i)(y - y_i)$$

Let $x = \bar{x}$, $y = \bar{y}$, and $w_i = \frac{1}{n}$, then

$$z_{xy} = \frac{1}{n} \sum_{i=1}^n (\bar{x} - x_i)(\bar{y} - y_i)$$

$$\text{COV}_{\text{biased}}(x, y) = 2.45635$$



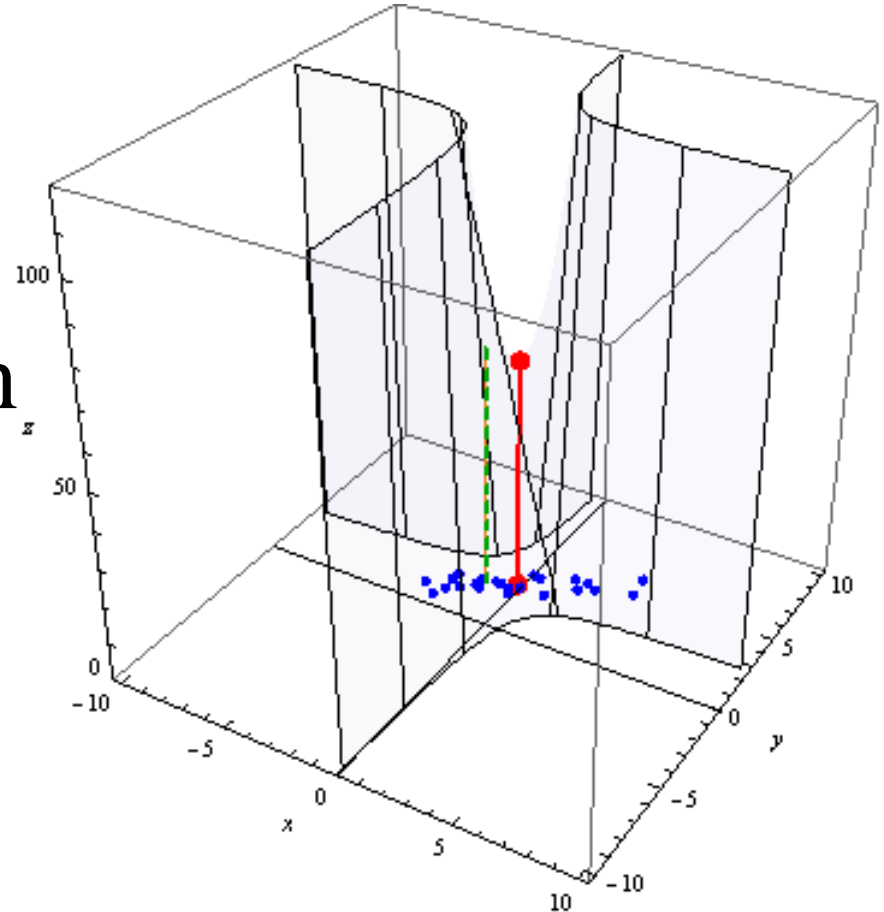
Sum of the Product of Differences: Unbiased Sample Covariance

$$z_{xy} = \sum_{i=1}^n w_i (x - x_i)(y - y_i)$$

Let $x = \bar{x}$, $y = \bar{y}$, and $w_i = \frac{1}{n-1}$, then

$$z_{xy} = \frac{1}{n-1} \sum_{i=1}^n (\bar{x} - x_i)(\bar{y} - y_i)$$

$$\text{COV}_{\text{unbiased}}(x, y) = 2.5587$$

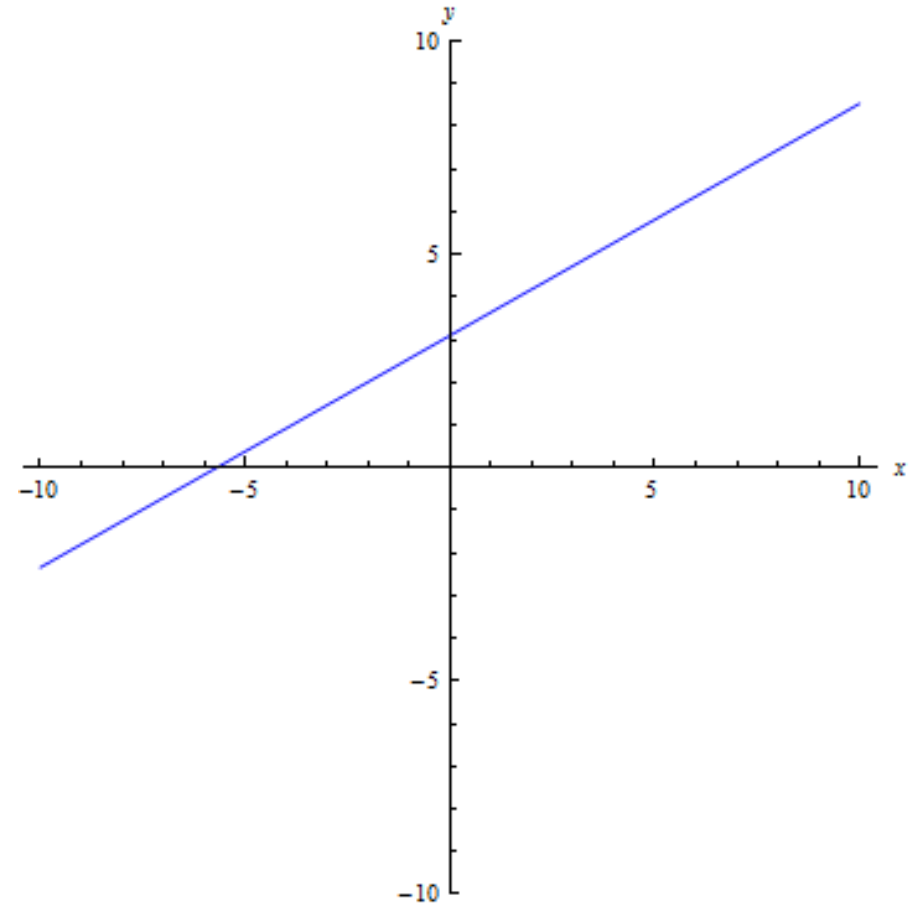


Line

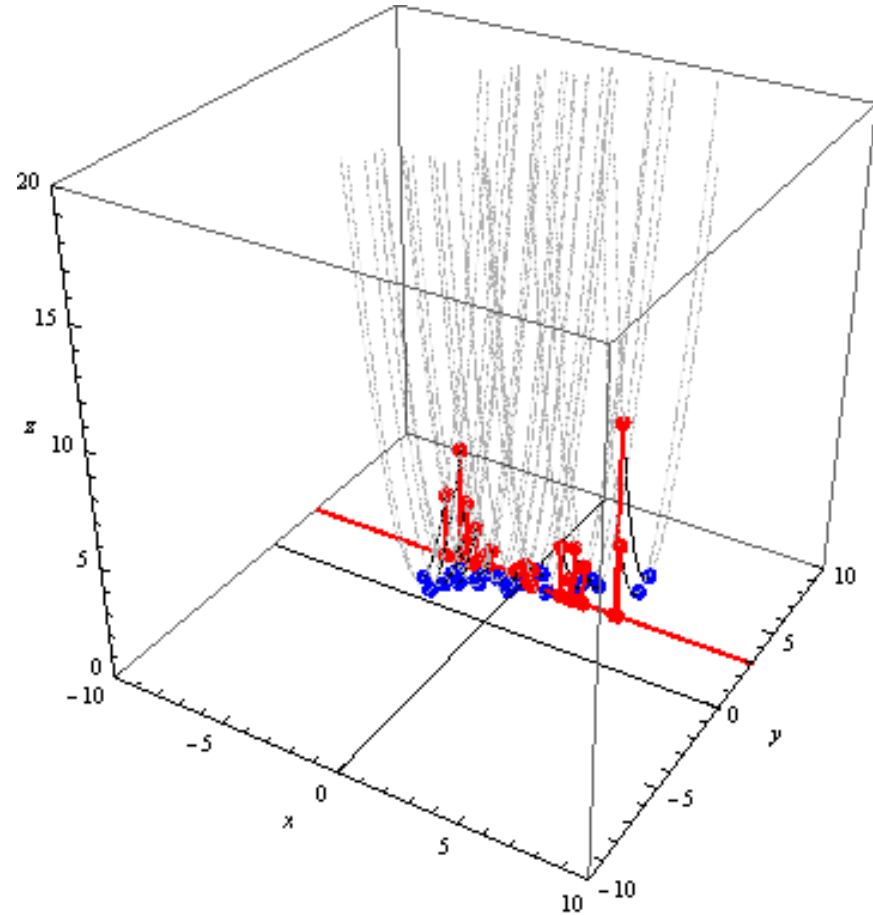
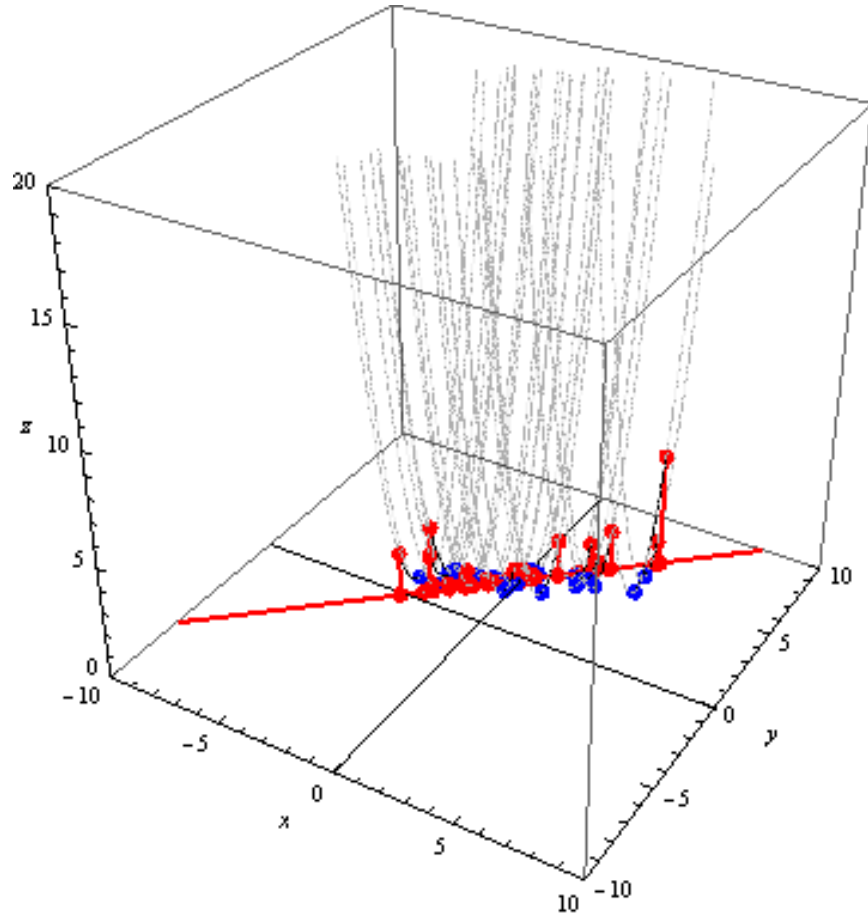
$$y = w(x - x_0) + y_0$$

$$y = wx - wx_0 + y_0$$

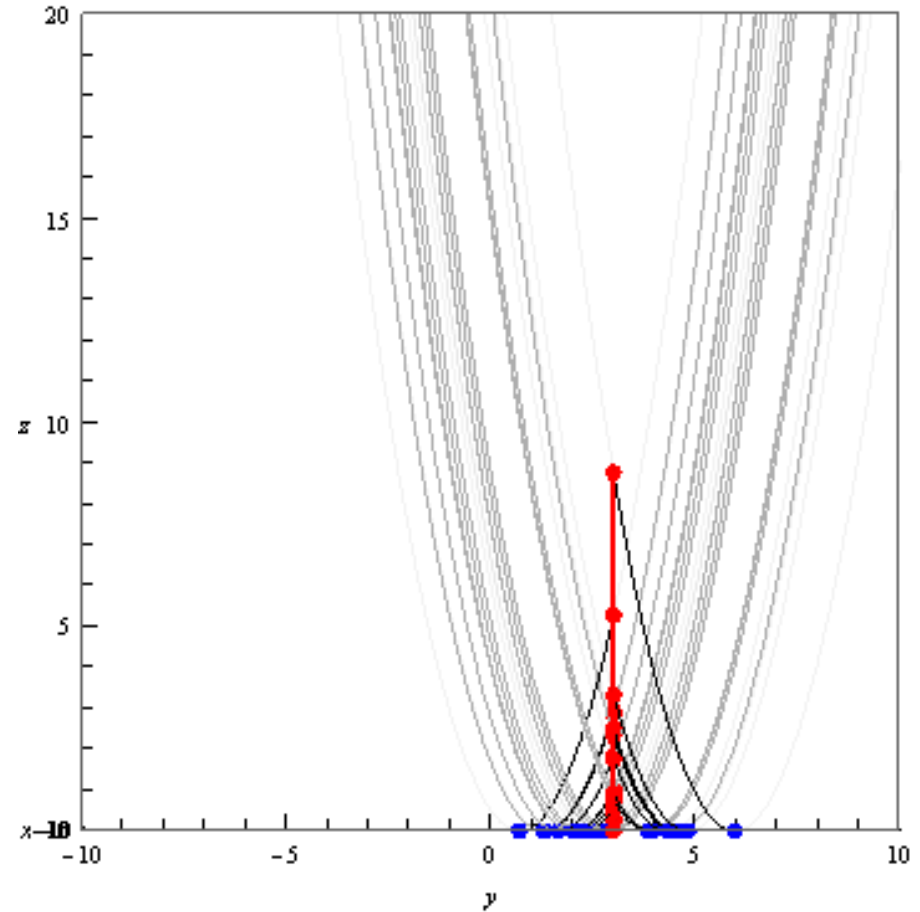
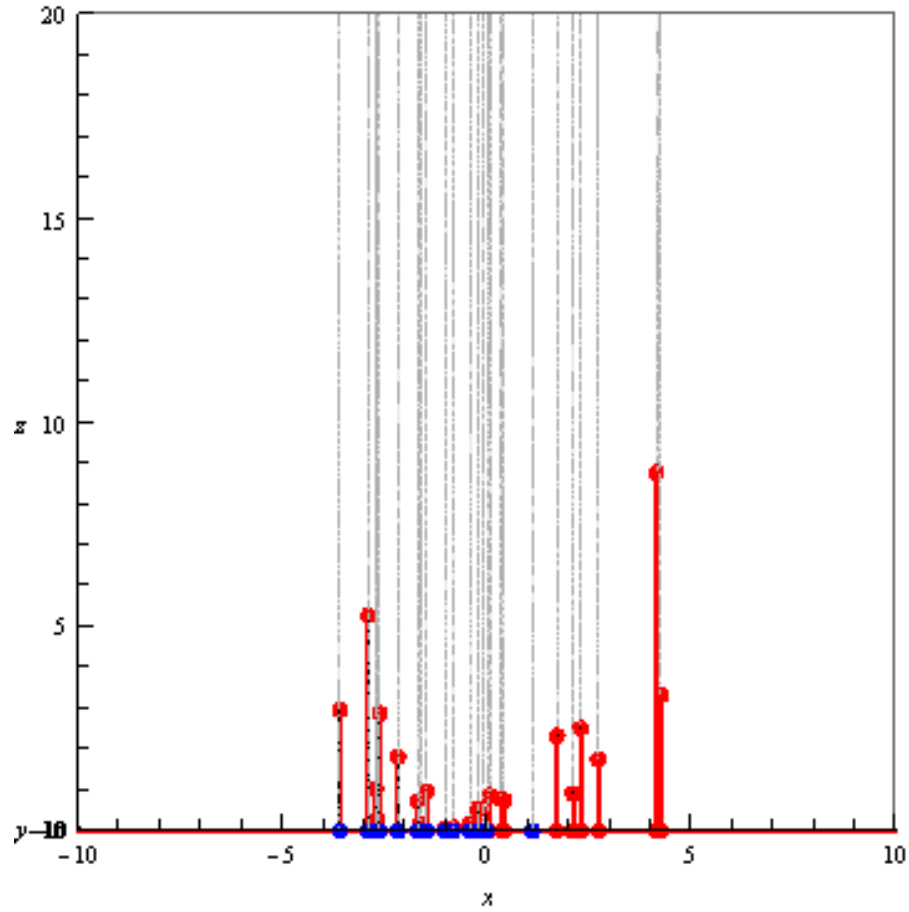
$$y = \underbrace{(w)}_m x + \underbrace{(-wx_0 + y_0)}_b$$



Linear Regression: Parabolas



Linear Regression: Parabolas

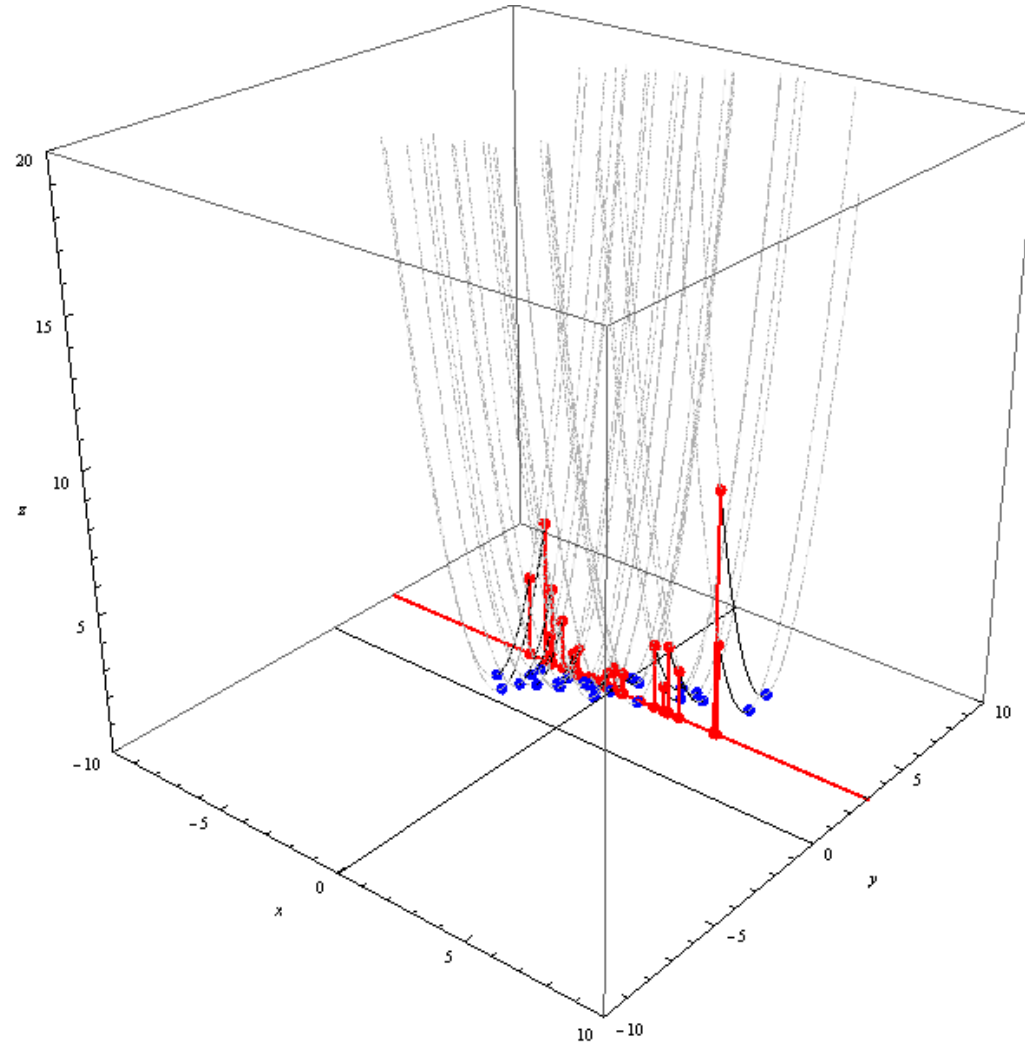


Linear Regression

$$y = w(x - \bar{x}) + \bar{y}$$

$$\begin{aligned} f(w) &= \sum_{i=1}^n \left((y_i) - (w(x_i - \bar{x}) + \bar{y}) \right)^2 \\ &= \sum_{i=1}^n \left(-(x_i - \bar{x})w + (y_i - \bar{y}) \right)^2 \end{aligned}$$

Linear Regression



Linear Regression

$$f(w) = \sum_{i=1}^n \left(-(x_i - \bar{x})w + (y_i - \bar{y}) \right)^2$$

$$\frac{df}{dw} = \sum_{i=1}^n 2 \left(-(x_i - \bar{x})w + (y_i - \bar{y}) \right) \left(-(x_i - \bar{x}) \right)$$

$$\frac{d^2 f}{dw^2} = \sum_{i=1}^n 2(x_i - \bar{x})^2$$

Minima: Scaled Variance & Covariance

$$\frac{df}{dw} = 0$$

$$\sum_{i=1}^n 2\left(-\left(x_i - \bar{x}\right)w + \left(y_i - \bar{y}\right)\right)\left(-\left(x_i - \bar{x}\right)\right) = 0$$

$$\sum_{i=1}^n \left(\left(x_i - \bar{x}\right)^2 w - \left(x_i - \bar{x}\right)\left(y_i - \bar{y}\right)\right) = 0$$

$$\sum_{i=1}^n \left(\left(x_i - \bar{x}\right)^2 w\right) - \sum_{i=1}^n \left(\left(x_i - \bar{x}\right)\left(y_i - \bar{y}\right)\right) = 0$$

Minima:

Sum of Squared Differences & Sum of the Product of Differences

$$\sum_{i=1}^n \left((x_i - \bar{x})^2 w \right) - \sum_{i=1}^n \left((x_i - \bar{x})(y_i - \bar{y}) \right) = 0$$

$$w \sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n \left((x_i - \bar{x})(y_i - \bar{y}) \right)$$

$$w = \frac{\sum_{i=1}^n \left((x_i - \bar{x})(y_i - \bar{y}) \right)}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

Linear Regression

$$\bar{x} = \frac{\sum_{i=1}^n x_i}{n} \quad \bar{y} = \frac{\sum_{i=1}^n y_i}{n} \quad w = \frac{\sum_{i=1}^n ((x_i - \bar{x})(y_i - \bar{y}))}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

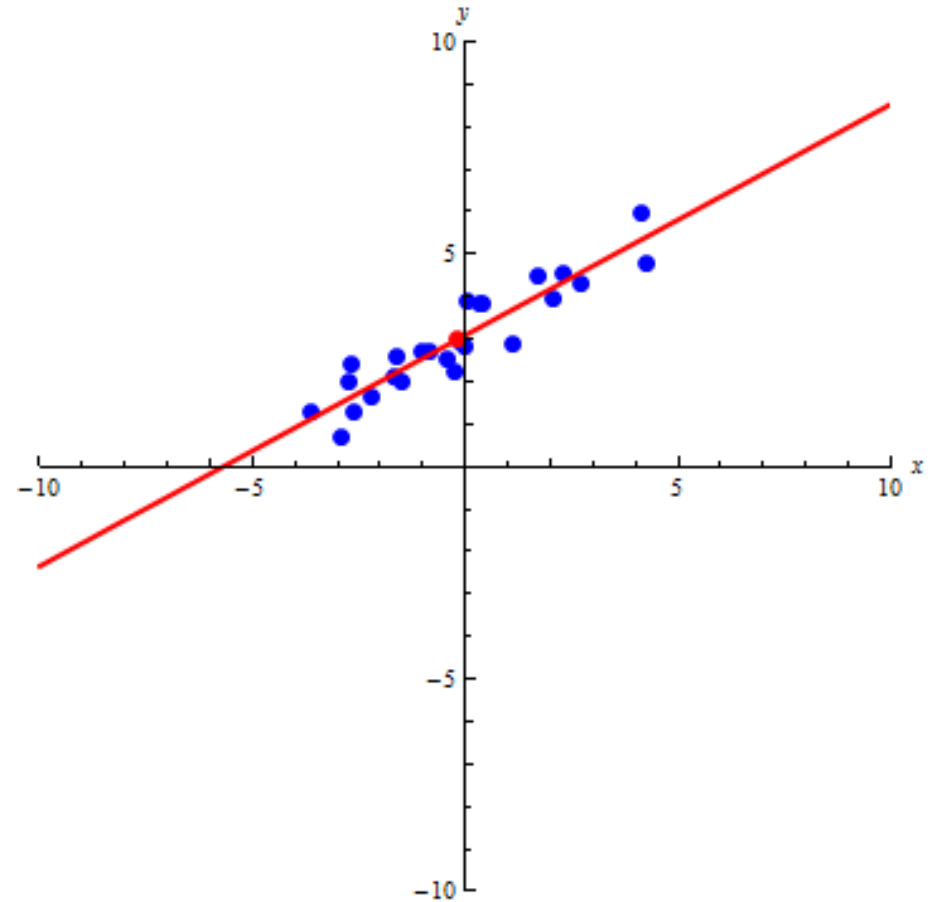
$$y = w(x - \bar{x}) + \bar{y}$$

$$y = \underbrace{(w)}_m x + \underbrace{(-w\bar{x} + \bar{y})}_b$$

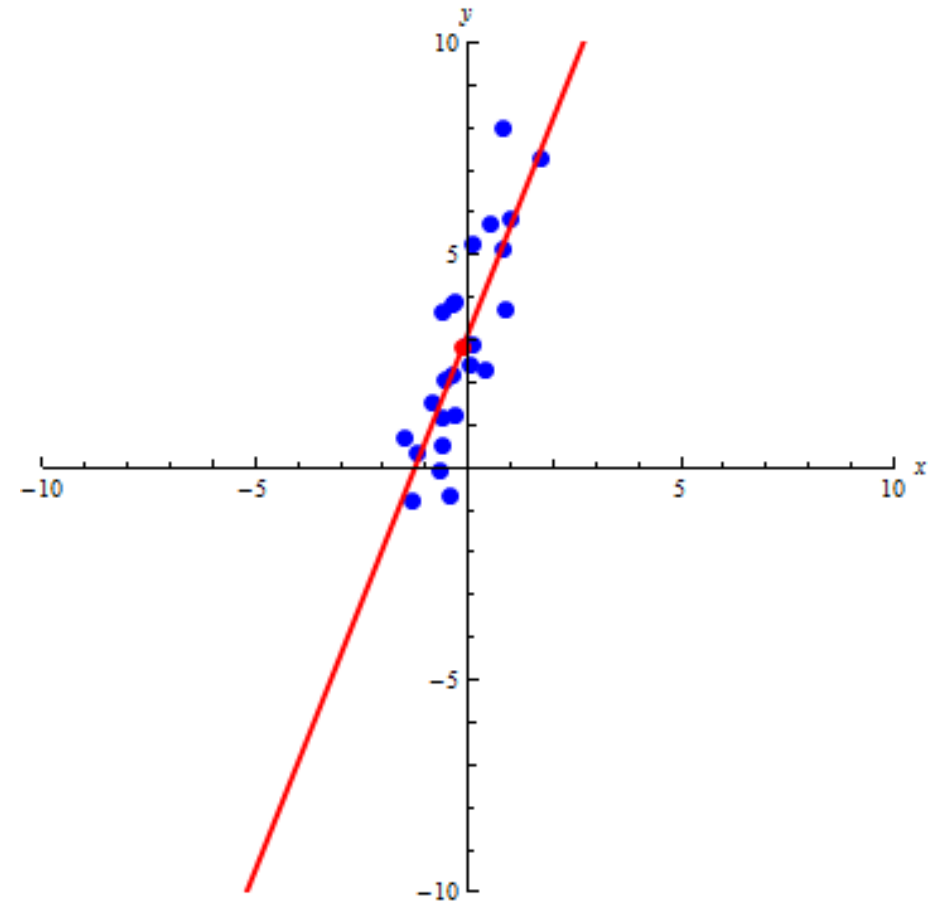
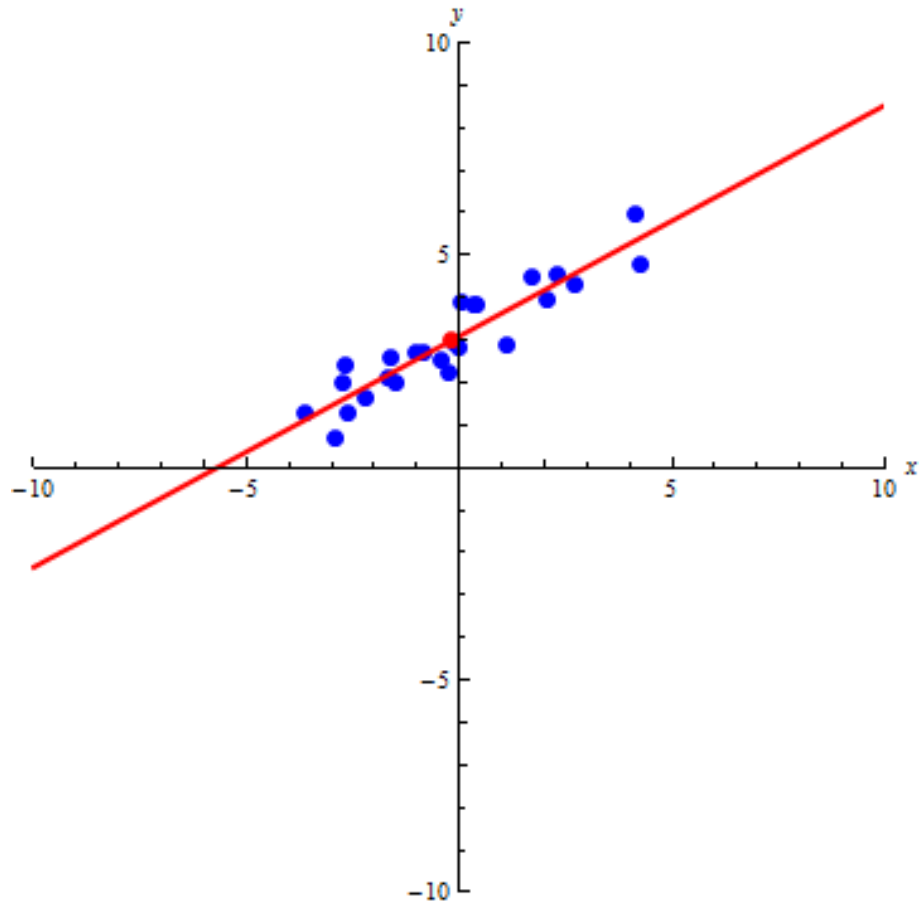
Linear Regression

- Slope: 0.543745
- Mean: $\{-0.208094, 2.98517\}$
- Equation:

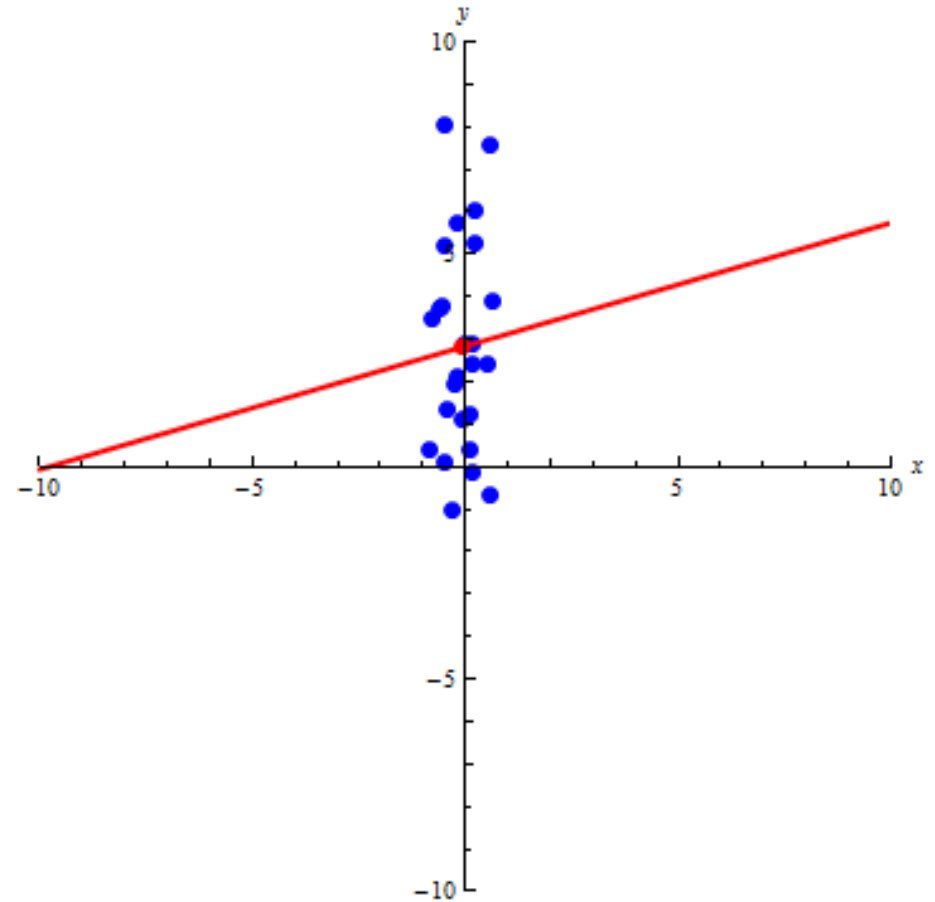
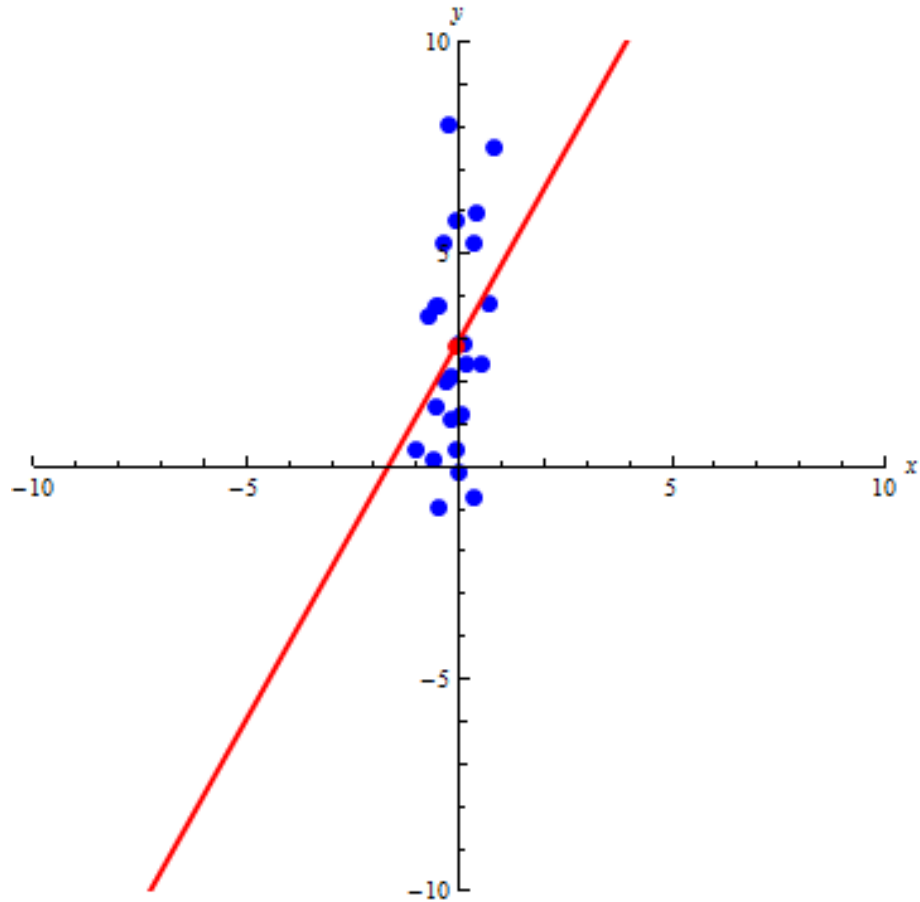
$$y = 0.543745x + 3.09832$$



Independent Variable Bias



Independent Variable Bias



Online Mean

$$\bar{x} = \frac{\sum_{i=1}^n x_i}{n}$$

$$\bar{x} = \frac{\sum_{i=1}^{n-1} x_i + x_n}{n}$$

$$\bar{x} = \frac{\sum_{i=1}^{n-1} x_i}{n} \cdot \frac{n-1}{n-1} + \frac{x_n}{n}$$

$$\bar{x}_n = \bar{x}_{n-1} \cdot \frac{n-1}{n} + \frac{x_n}{n}$$

$$\bar{x} = \frac{\sum_{i=1}^{n-1} x_i}{n} + \frac{x_n}{n}$$

$$\bar{x} = \frac{\sum_{i=1}^{n-1} x_i}{n-1} \cdot \frac{n-1}{n} + \frac{x_n}{n}$$

$$\bar{x}_n = \frac{\bar{x}_{n-1} (n-1) + x_n}{n}$$

One Way To Program Simple Linear Regression

```
/*
Filename: main.cpp
To compile and run on linprog4.cs.fsu.edu: g++47 -o main.exe main.cpp -std=c++11 -O3 -Wall -Wextra -Werror -static && ./main.exe
*/
#include <iostream>

template <size_t n> std::ostream & operator << ( std::ostream & o, char const ( & c ) [ n ] ) { for ( size_t i = 0; i < n - 1; ++i ) std::cout << c [ i ]; return o; }
template <typename T, size_t n > std::ostream & operator << ( std::ostream & o, T const ( & v ) [ n ] ) {
    auto i = std::begin ( v ); auto end = std::end ( v ); o << '{'; if ( i != end ) { o << *i; for ( ++i; i != end; ++i ) o << ', ' << *i; } o << '}' ; return o;
}

int main () {
    using number = double;
    enum dimensions { X, Y, DIMENSIONS };
    number points [ ] [ DIMENSIONS ] {
        {-2.70238, 2.43155}, {-2.18612, 1.63634}, {0.405141, 3.84976}, {-1.03072, 2.7144}, {1.10033, 2.89639},
        {-1.63906, 2.56916}, {2.27983, 4.57127}, {-0.836348, 2.70824}, {-2.90988, 0.685828}, {-0.104817, 2.95222},
        {-0.226538, 2.24849}, {-2.64364, 1.28981}, {-0.00953108, 2.84022}, {0.336282, 3.86626}, {4.12633, 5.94993},
        {1.70053, 4.51368}, {-1.4793, 1.99986}, {0.0467884, 3.91811}, {-1.70285, 2.12758}, {-3.61035, 1.26436},
        {2.08504, 3.94459}, {4.23512, 4.80965}, {2.70993, 4.30984}, {-2.72741, 1.97363}, {-0.418723, 2.55797}
    };
    number mean [ DIMENSIONS ] { 0 };
    size_t n { 1 };
    for ( auto & point : points ) {
        mean [ X ] = ( mean [ X ] * ( n - 1 ) + point [ X ] ) / n;
        mean [ Y ] = ( mean [ Y ] * ( n - 1 ) + point [ Y ] ) / n;
        ++n;
    }
    std::cout << "mean: " << mean << '\n';
    number residuals [ DIMENSIONS ] { 0 };
    number sum_of_product_of_differences_xy { 0 };
    number sum_of_squared_differences_xx { 0 };
    for ( auto & point : points ) {
        residuals [ X ] = point [ X ] - mean [ X ];
        residuals [ Y ] = point [ Y ] - mean [ Y ];
        sum_of_product_of_differences_xy += residuals [ X ] * residuals [ Y ];
        sum_of_squared_differences_xx += residuals [ X ] * residuals [ X ];
    }
    number w { sum_of_product_of_differences_xy / sum_of_squared_differences_xx };
    std::cout << "w: " << w << '\n';
    number m { w };
    number b { - w * mean [ X ] + mean [ Y ] };
    std::cout << "y = " << m << " * x + " << b << '\n';
}
```

Conclusions

- The sum of parabolas is a parabola.
- Parabolas produce a global minima.
- Parabolas can be used to produce Means and Linear Regressions from Variances and Covariances.
- As presented, Linear Regression is biased toward the independent variable; moreover, the type of Linear Regression cannot produce a line that is straight up and down regardless of the data.
- The Online Mean can be used to calculate the mean of an unknown amount of data or the mean of continuously generated data.