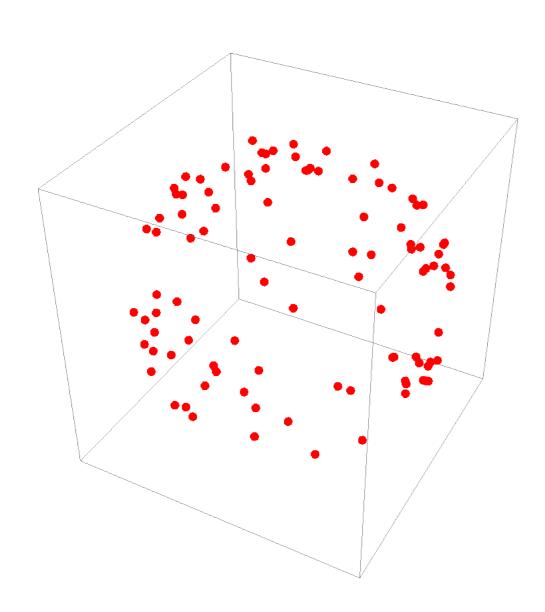
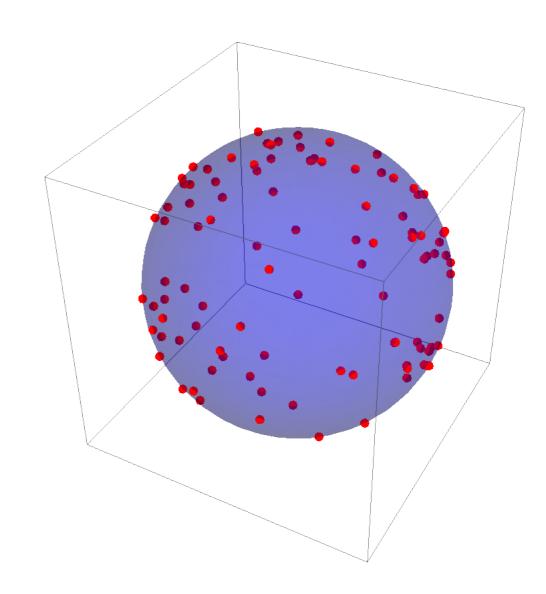
Sphere Fitting

Part I

Marcelo Ferreira Siqueira

What's Sphere Fitting?





Why Sphere Fitting?

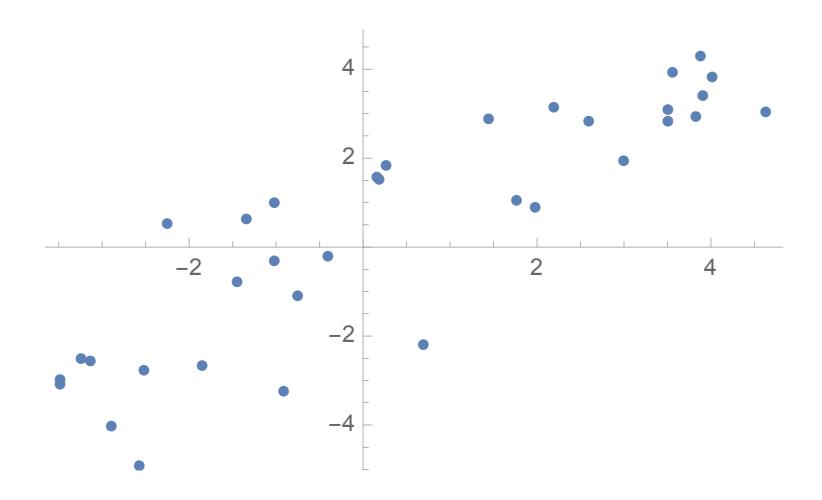
How do we decide what a best fit is?

Digression:

fit to a line using least-squares

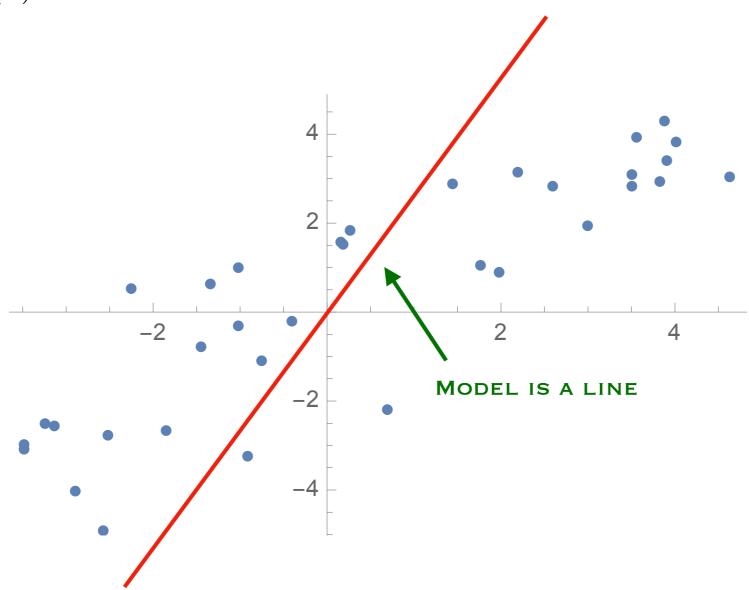
Data:

$$\{(x_i, y_i)\}_{i=1}^n$$

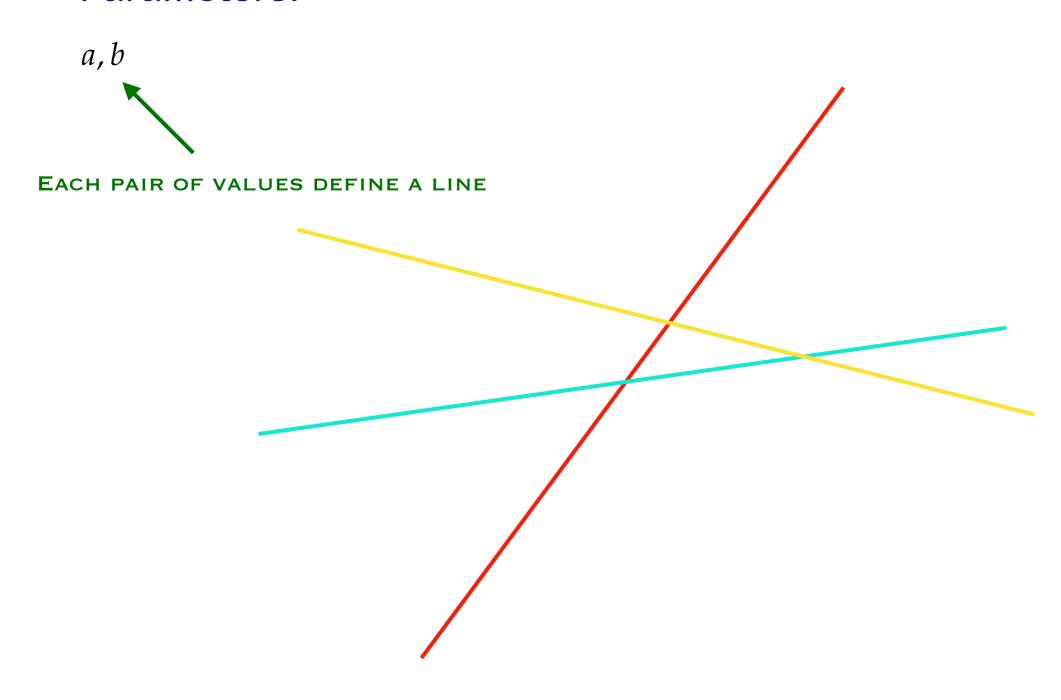


Model:

$$f: \mathbb{R} \to \mathbb{R}, \quad f(x) = ax + b$$

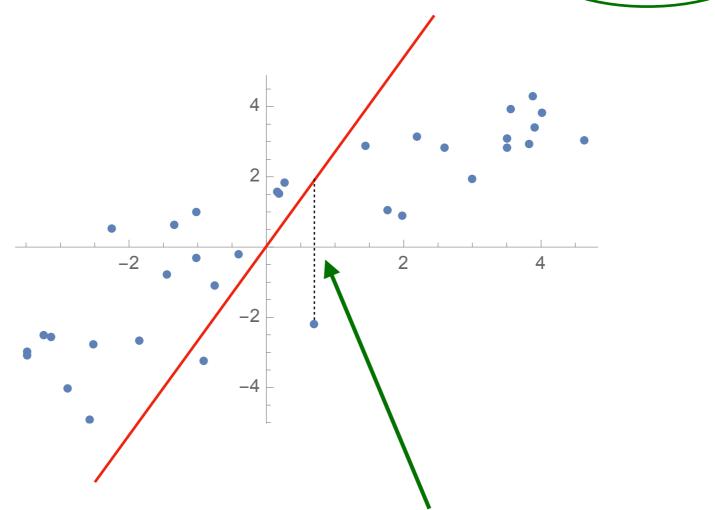


Parameters:



Define a residual function for each data point:

$$g_i: \mathbb{R}^2 \to \mathbb{R}$$
 such that $g_i(a,b) = f(x_i) - y_i = ax_i + b - y_i$



RESIDUAL IS SIMPLY A "VERTICAL" SIGNED DISTANCE TO THE LINE

Define a functional as the sum of squares of residuals:

$$G: \mathbb{R}^2 \to \mathbb{R}$$
 such that $G(a,b) = \sum_{i=1}^n (g_i(a,b))^2$

IT IS A QUADRATIC FUNCTION FOR THE CASE OF A LINE

Solution:

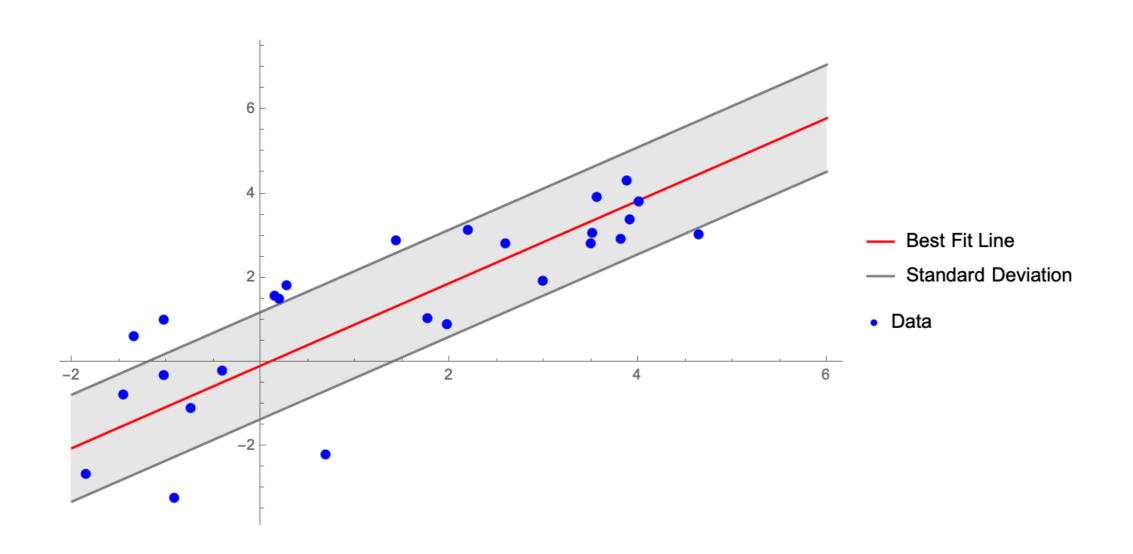
 $\operatorname{argmin}_{(a,b)\in\mathbb{R}^2}\{G(a,b)\}$

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MAY NOT BE UNIQUE!

Solution:

$$\operatorname{argmin}_{(a,b)\in\mathbb{R}^2}\{G(a,b)\}$$



This is an instance of the classic linear least-squares problem:

$$\begin{cases} ax_1 + b - y_1 &= 0 \\ ax_2 + b - y_2 &= 0 \\ \vdots & \vdots & \\ ax_n + b - y_n &= 0 \end{cases} \implies \begin{pmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} - \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \implies Xc - y = 0$$

We need to solve

$$\operatorname{argmin}_{c \in \mathbb{R}^2} \{ \|Xc - y\|_2 \}$$

but we actually solve

$$\operatorname{argmin}_{c \in \mathbb{R}^2} \left\{ \|Xc - y\|_2^2 \right\}$$
 (which is equivalent to $\operatorname{argmin}_{(a,b) \in \mathbb{R}^2} \left\{ G(a,b) \right\}$)

Consider the problem of fitting a circle to a set of planar points

Let us define the functional as

$$G(r, x_c, y_c) = \sum_{i=1}^{n} (\pi r^2 - \pi ((x_i - x_c)^2 + (y_i - y_c)^2))^2$$

AREA OF A CIRCLE OF RADIUS R

Observe that the residual functions are:

$$g_i(r, x_c, y_c) = \pi r^2 - \pi ((x_i - x_c)^2 + (y_i - y_c)^2)$$

Let

$$J(r, x_c, y_c) = \frac{G(r, x_c, y_c)}{\pi^2} = \sum_{i=1}^n (r^2 - ((x_i - x_c)^2 + (y_i - y_c)^2))^2$$

So,

$$\frac{\partial J}{\partial r}(r, x_c, y_c) = 4 \left[\sum_{i=1}^{n} \left(r^2 - ((x_i - x_c)^2 + (y_i - y_c)^2) \right) \right] r$$

$$\frac{\partial J}{\partial x_c}(r, x_c, y_c) = 4 \left[\sum_{i=1}^n \left(r^2 - ((x_i - x_c)^2 + (y_i - y_c)^2) \right) \right] (x_c - x_i)$$

$$\frac{\partial J}{\partial y_c}(r, x_c, y_c) = 4 \left[\sum_{i=1}^n \left(r^2 - ((x_i - x_c)^2 + (y_i - y_c)^2) \right) \right] (y_c - x_i)$$

We want to find

$$(r, x_c, y_c) \in \mathbb{R}^3$$

such that

$$\frac{\partial J}{\partial r}(r, x_c, y_c) = \frac{\partial J}{\partial x_c}(r, x_c, y_c) = \frac{\partial J}{\partial y_c}(r, x_c, y_c) = 0$$

This can lead us to the classic linear least-squares problem, but...

From

$$\frac{\partial J}{\partial r}(r, x_c, y_c) = 0$$

we get

$$nr^{2} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right)$$

From

$$\frac{\partial J}{\partial x_c}(r, x_c, y_c) = 0$$

we get

$$\sum_{i=1}^{n} \left(r^2 - \left((x_i - x_c)^2 + (y_i - y_c)^2 \right) \right) x_i = \sum_{i=1}^{n} \left(r^2 - \left((x_i - x_c)^2 + (y_i - y_c)^2 \right) \right) x_c$$

From

$$nr^{2} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right)$$

and

$$\sum_{i=1}^{n} \left(r^2 - \left((x_i - x_c)^2 + (y_i - y_c)^2 \right) \right) x_i = \sum_{i=1}^{n} \left(r^2 - \left((x_i - x_c)^2 + (y_i - y_c)^2 \right) \right) x_c$$

we get

$$r^{2} \sum_{i=1}^{n} x_{i} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right) x_{i}$$

Similarly,

$$r^{2} \sum_{i=1}^{n} y_{i} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right) y_{i}$$

Using the notation

$$\sum_{x} \sum_{i=1}^{n} x_{i} \qquad \sum_{y} \sum_{i=1}^{n} y_{i} \qquad \sum_{x^{2}} \sum_{i=1}^{n} x_{i}^{2} \qquad \sum_{y^{2}} \sum_{i=1}^{n} y_{i}^{2}$$

$$\sum_{xy} \sum_{i=1}^{n} x_{i}y_{i} \qquad \sum_{x^{3}} \sum_{i=1}^{n} x_{i}^{3} \qquad \sum_{y^{3}} \sum_{i=1}^{n} y_{i}^{3} \qquad \sum_{x^{2}y} \sum_{i=1}^{n} x_{i}^{2}y_{i}$$

$$\sum_{xy^{2}} \sum_{i=1}^{n} x_{i}y_{i}^{2}$$

equation

$$nr^{2} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right)$$

becomes

$$nr^{2} = \sum_{x^{2}} -2\sum_{x} x_{c} + nx_{c}^{2} + \sum_{y^{2}} -2\sum_{y} y_{c} + ny_{c}^{2}$$

and equation

$$r^{2} \sum_{i=1}^{n} x_{i} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right) x_{i}$$

becomes

$$r^{2} \sum_{x} = \sum_{x^{3}} -2 \sum_{x^{2}} x_{c} + \sum_{x} x_{c}^{2} + \sum_{xy^{2}} -2 \sum_{xy} y_{c} + \sum_{x} y_{c}^{2}$$

and equation

$$r^{2} \sum_{i=1}^{n} y_{i} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right) y_{i}$$

becomes

$$r^{2} \sum_{y} = \sum_{x^{2}y} -2 \sum_{xy} x_{c} + \sum_{y} x_{c}^{2} + \sum_{y^{3}} -2 \sum_{y^{2}} y_{c} + \sum_{y} y_{c}^{2}$$

Multiply

$$nr^{2} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right)$$

by

$$\sum_{\chi}$$

and subtract

$$n(r^{2} \Sigma_{x}) = n\left(\Sigma_{x^{3}} - 2\sum_{x^{2}} x_{c} + \sum_{x} x_{c}^{2} + \sum_{xy^{2}} -2\sum_{xy} y_{c} + \sum_{x} y_{c}^{2}\right)$$

to give

$$\sum_{x^2} \sum_{x} -n \sum_{x^3} -2x_c \left((\sum_{x})^2 - n \sum_{x^2} \right) + \sum_{x} \sum_{y^2} -n \sum_{xy^2} -2y_c \left(\sum_{x} \sum_{y} -n \sum_{xy} \right) = 0$$

Multiply

$$nr^{2} = \sum_{i=1}^{n} \left((x_{i} - x_{c})^{2} + (y_{i} - y_{c})^{2} \right)$$

by

$$\sum_{y}$$

and subtract

$$n(r^{2} \Sigma_{y}) = n(\Sigma_{x^{2}y} - 2 \Sigma_{xy} x_{c} + \Sigma_{y} x_{c}^{2} + \Sigma_{y^{3}} - 2 \Sigma_{y^{2}} y_{c} + \Sigma_{y} y_{c}^{2})$$

to give

$$\sum_{x^2} \sum_{y} -n \sum_{x^2 y} -2x_c (\sum_{x} \sum_{y} -n \sum_{xy}) + \sum_{y} \sum_{y^2} -n \sum_{y^3} -2y_c \left(\left(\sum_{y} \right)^2 -n \sum_{y^2} \right) = 0$$

Now we solve a linear system of 2 equations and 2 unknowns:

$$\left(\begin{array}{cc} a_{11} & a_{12} \\ a_{21} & a_{22} \end{array}\right) \left(\begin{array}{c} x_c \\ y_c \end{array}\right) = \left(\begin{array}{c} b_1 \\ b_2 \end{array}\right),$$

where

$$a_{11} = 2\left((\sum_{x})^{2} - n\sum_{x^{2}}\right), \qquad a_{12} = 2\left(\sum_{x}\sum_{y} - n\sum_{xy}\right)$$

$$a_{21} = 2\left(\sum_{x}\sum_{y} - n\sum_{xy}\right), \qquad a_{22} = 2\left(\left(\sum_{y}\right)^{2} - n\sum_{y^{2}}\right)$$

$$b_{1} = \sum_{x^{2}}\sum_{x} - n\sum_{x^{3}} + \sum_{x}\sum_{y^{2}} - n\sum_{xy^{2}}, \quad b_{2} = \sum_{x^{2}}\sum_{y} - n\sum_{y^{3}} + \sum_{y}\sum_{y^{2}} - n\sum_{x^{2}y}$$

Once we obtain the coordinates of the center, the radius is given as:

$$r^{2} = \frac{1}{n} \left(\sum_{x^{2}} 2 \sum_{x} x_{c} + n x_{c}^{2} + \sum_{y^{2}} -2 \sum_{y} y_{c} + n y_{c}^{2} \right)$$

This approach is credited to:

Samuel M. Thomas and Y. T. Chan

A simple approach for the estimation of circular arc center and radius, *Computer Vision*, *Graphics*, and *Image Processing* (CVGIP), 45(3), March, p. 362-370, 1989.

Its extension to the 3d case is trivial and can be found at:

Y. D. Sumith

Fast geometric fit algorithm for sphere using exact solution, ArXiv, 2015.

HTTPS://DBLP.ORG/DB/JOURNALS/CORR/CORR1506.HTML#YD15

Experiment

Considered a sphere of radius 5 centered at (0,0,0)

Generated 100 points uniformly distributed over the sphere

Added white noise to the points using standard deviation equal to

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• 0.5000 (~10.00% of the radius)
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- 0.0500 (~ 1.00% of the radius)
- 0.0050 (~ 0.10% of the radius)
- 0.0005 (~ 0.01% of the radius)

Ran algorithm for computing center and radius from the points

Results

STDDEV	RADIUS	CENTER	SoSR
0.5000	5.10152	(+0.76577800, -0.01023420, -0.12833200)	2990.160000
0.0500	4.99944	(-0.01199920, +0.00508314, +0.01469530)	26.48840000
0.0050	5.00016	(-0.00018245, -0.00057873, -0.00375501)	0.254553000
0.0005	5.00006	(+0.00001887, +0.00012307, +0.00008819)	0.002555000

Implementation

C++ 17 and Eigen 3.3.9

Unit tests

Final Remarks

Algebraic vs Geometric Fitting (difference in the residuals):

