Technical Computing for the Earth Sciences, Lecture 2:

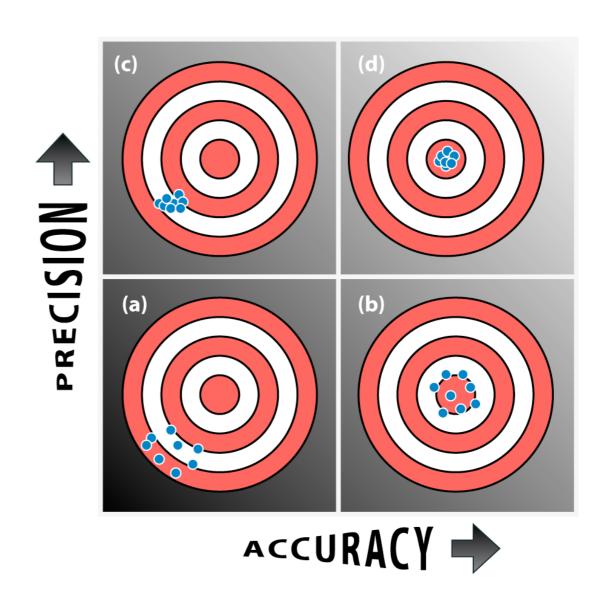
Data and uncertainty Datasets as vectors and matrices

EARS 80.03

Uncertainty

What is uncertainty?

An estimate of "distance from some true or central value"?



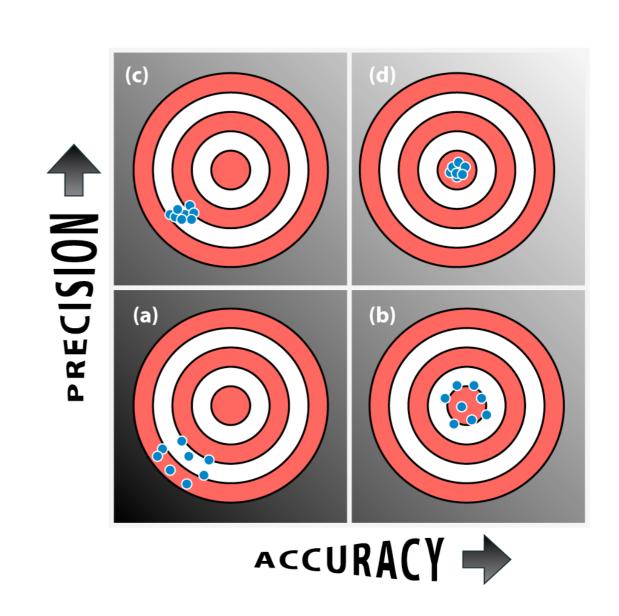
Uncertainty

What is uncertainty?

An estimate of "distance from some true or central value"?

- Accuracy: distance from true value
- Precision: distance from central value (scatter)
- Can't always determine accuracy, but can always determine precision

First, some terminology around vectors and matrices



Vectors

A vector is a one-dimensional array of numbers

(or as B & V put it, "an ordered finite list of numbers")

e.g.:

 $\begin{bmatrix}
 -1.1 \\
 0.0 \\
 3.6 \\
 -7.2
 \end{bmatrix}$

 A vector of length n can be thought of as a point in n-dimensional space

Matrices

A matrix is a two-dimensional array of numbers

$$\begin{bmatrix} 0 & 1 & -2.3 & 0.1 \\ 1.3 & 4 & -0.1 & 0 \\ 4.1 & -1 & 0 & 1.7 \end{bmatrix}$$

- Have rows (horizontal) and columns (vertical)
- Matrices have various interpretations. Can be thought of:
 - as just a bunch of vectors
 - as systems of linear equations
 - as functions that act on vectors
 - etc.

Can be multiplied

$$\begin{bmatrix}
 -1.1 \\
 0.0 \\
 3.6 \\
 -7.2
 \end{bmatrix}$$

$$\left[\begin{array}{ccccc}
0 & 1 & -2.3 & 0.1 \\
1.3 & 4 & -0.1 & 0 \\
4.1 & -1 & 0 & 1.7
\end{array}\right]^{*}$$

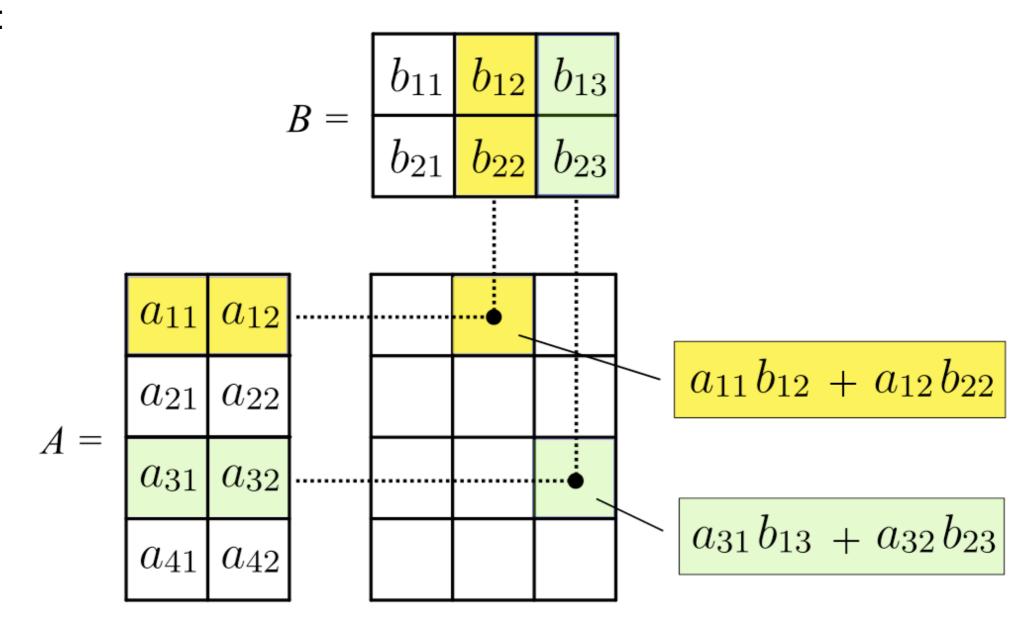
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$$\begin{bmatrix} 0 & 1 & -2.3 & 0.1 \\ 1.3 & 4 & -0.1 & 0 \\ 4.1 & -1 & 0 & 1.7 \end{bmatrix}$$
*Ox-1.1+1*0 + -2.5*3.6 +0.1 x - 7.2

etc.

Can be multiplied



Can be multiplied

e.g.:
$$\begin{bmatrix} -1.1 \\ 0.0 \\ 3.6 \\ -7.2 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 & -2.3 & 0.1 \\ 1.3 & 4 & -0.1 & 0 \\ 4.1 & -1 & 0 & 1.7 \end{bmatrix}$$
 * -9 -1.7 -16.75

- Matrix multiplication isn't commutative
- Have to be right size (# columns in left matrix have to equal # rows in right matrix)
- Try it out in Julia to see what does and doesn't work

Uncertainty

How do we measure uncertainty?

"Distance from some true or central value"?

Two problems:

- A. How do we measure distance?
- B. What should we use as the true or central value?

A: How do we measure distance?

Norms

$$|\mathbf{x}| = \left(\sum_{i=1}^{n} |x_i|^p\right)^{1/p}$$

Examples:

$$a = \begin{bmatrix} 3 \\ 0 \end{bmatrix} \qquad b = \begin{bmatrix} 0 \\ -4 \end{bmatrix}$$

$$a = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

$$\mathcal{X} = \frac{\mathbf{0}}{\mathbf{0}} - \frac{\mathbf{0}}{\mathbf{0}} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

$$p = 0$$

- Technically not valid
- Sometimes said to be # of non-zero dimensions of **x** (i.e., # of dimensions in which a and b differ)

$$p = 1$$

- = $\Sigma |x_i|$
- "Manhattan norm" or "taxicab norm"

$$p = 2$$

- = $(\Sigma |x_i|^2)^{1/2}$
- "Euclidian norm" or "Euclidian distance"

• Largest element of **x**

A: How do we measure distance?

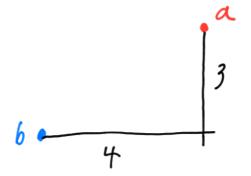
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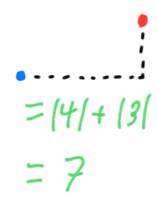


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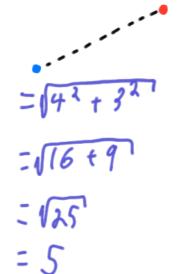
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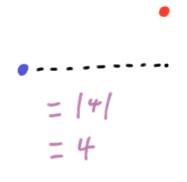
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- = $(\Sigma |x_i|^2)^{1/2}$
- "Euclidian norm" or "Euclidian distance"



$$\lim p \to \infty$$

Largest element of x



No interesting norms for 2 ?So it seems. Possibly related: this Let's take off the outer $\wedge(1/p)$ just to make this a bit simpler

Discrepancy

Say we have some dataset **D** containing a whole bunch of data points *Let's assume these points (vectors) are one-dimensional (i.e, just numbers) to make this easier to think about

What is the value \mathbf{m} that minimizes the discrepancy of \mathbf{D} - \mathbf{m} , i.e. $\Sigma |\mathbf{D}_i - \mathbf{m}|^p$?

$$p = 0$$

Want a point **m** that...

 Minimizes the number of times that one of the points in **D** is not equal to **m**

*If we're willing to take the slightly non-standard position that $0^0 = 0$

$$p = 1$$

Want a point **m** that...

 Minimizes the average distance between **m** and each point in **D**

$$p = 2$$

Want a point **m** that...

 Minimizes the average squared distance between
 m and each point in D Let's take off the outer $\wedge(1/p)$ just to make this a bit simpler

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Say
$$\mathbf{D} = [1,1,2,3,8]$$

$$p = 0$$

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times that one of the points

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What should we use as the metric for distance and for central value when measuring scatter?

- p=1 norm, p=1 central value (median)?
 - -> median absolute deviation
 - = $median(abs.(\mathbf{x} .- median(\mathbf{x})))$

- p=2 norm, p=2 central value (mean)?
 - -> standard deviation

$$= \left(\frac{1}{N-1} \sum_{i=1}^{N} (x_i - \mu)^2\right)^{\frac{1}{2}}$$
 where μ is the mean of \mathbf{x} .

- In Julia: $sqrt(sum((\mathbf{x} .- mean(\mathbf{x})).^2)/(length(\mathbf{x})-1))$, or just $std(\mathbf{x})$
- almost the same as $sqrt(mean((x .- mean(x)).^2))$, if it weren't for the -1 in N-1 (which is to do with sample vs population statistics)*

Some useful functions for working with data and uncertainty

```
julia> rand(dims) # Uniform random numbers between 0 and 1 julia> randn(dims) # Gaussian random numbers with \mu=0, \sigma=1
```

In stdlibs:

```
julia> using Statistics, StatsBase
julia> mean(data)
julia> std(data) # Standard deviation
julia> median(data)
julia> percentile(data)
```