

Information on this module

Instructor Federico Poloni, federico.poloni@unipi.it. Office 359 DN.

Office hours (ricevimento): by appointment.

Lectures Alternating with Antonio. Irregular schedule, especially at the beginning.

Content

Antonio's part: solve **all** the problems: $\min_{\mathbf{x} \in S} f(\mathbf{x})$.

This part: solve a **very specific** problem:

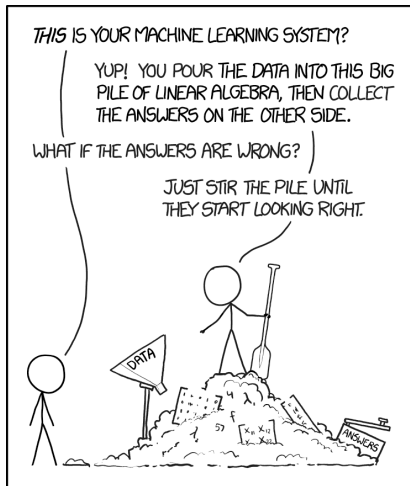
$$\min_{\mathbf{x} \in \mathbb{R}^n} \|\mathbf{Ax} - \mathbf{y}\|_2. \quad (\text{LS})$$

“Solve a linear system, or find the best approximation of the solution in a very specific norm”

Why is this still relevant? It is a problem we can understand well:

- ▶ Specialized algorithms that compute solutions quickly and with very high accuracy, even for very large dimensions (think $\approx 1M \times 1M$ matrices);
- ▶ We are **very** demanding: can we get the solution up to machine precision 10^{-16} ? If no, why?
- ▶ Often appears as sub-step in more difficult problems;
- ▶ Useful for theoretical understanding: e.g., principal components.

Obligatory XKCD



<https://xkcd.com/1838/>

Books

Books (for this part)

1. Trefethen–Bau, *Numerical Linear Algebra*. **Recommended**: we will follow its approach.
 2. Demmel, *Applied Numerical Linear Algebra*. Alternative source for other explanations, exercises. . .
 3. Strang, *Linear Algebra and Learning from Data*, <https://epubs.siam.org/doi/book/10.1137/1.9780692196380>. Another alternate source to review things from a different approach. Not very detailed with proofs.
 4. Eldén, *Matrix Methods in Data Mining and Pattern Recognition*, <https://epubs.siam.org/doi/book/10.1137/1.9780898718867>. Not very detailed for our purposes, but gives good insight on some topics.
2. and 3. are accessible for free from within our university network (or via VPN <https://start.unipi.it/en/help-ict/vpn/>).

Languages

Matlab: proprietary language specialized in matrix computations. Clunky language with a big Visual Basic-like IDE, great libraries and syntactic sugar for numerics.

Python: great simple language with decent numerics libraries stapled onto it.

C/C++, Fortran (yes, that's still a thing): sometimes needed for best performance.

Julia: newer, same age as Go / Rust. Stable, but tooling still not perfect. It tries to combine all advantages: syntactic sugar for numerics + libraries + fast loops.

Under the core: the same libraries (**Blas/Lapack**) for basic operations: summing vectors, multiplying matrices. . .

How to install Matlab

Available without charge to Unipi students:

- ▶ Go to <https://unipi.it/matlab>;
- ▶ Log on with your @studenti.unipi.it account;
- ▶ Create a Mathworks account;
- ▶ Download and install.

Suggested to save disk space and bandwidth: install only the toolboxes that you need. For this course: **Matlab, Symbolic Toolbox, Optimization Toolbox**.

You may be interested also in: Statistics and Machine Learning Toolbox, Neural Network Toolbox.

Alternative: **Cloud Matlab** <https://matlab.mathworks.com/>
(with the same account).

Matlab introduction

Command window

```
>> a = 19
```

```
a =
```

```
19
```

```
>> b = 5
```

```
b =
```

```
5
```

```
>> a+b
```

```
ans =
```

```
24
```

```
>> c = 2*a + b; % semicolons suppress output
```

Accuracy of operations

Default data type: double (IEEE standard binary64).

Warning: operations are only accurate 'up to 16 digits'.

```
>> (1/98) * 98 - 1  
ans =  
-1.1102e-16
```

(exponential notation: stands for -1.1102×10^{-16})

Understanding the effect of these approximations will be an important part of this module.

Numerical linear algebra people are very demanding: we have “simple” problems, and we wish to solve them **up to 10^{-16}** , or something around there. Not 10^{-8} or 10^{-12} .