

Introduction

Lecture "Mathematics of Learning (Maths of Data Science 1)"

Andreas Bärmann Friedrich-Alexander-Universität Erlangen-Nürnberg Winter semester 2022





- These lecture notes are based on the material by Frauke Liers from the summer semester 2022. It is a compilation from different sources that will be mentioned for further references during the lecture.
- Among the different sources: Many thanks to Frauke Liers, Martin Burger, Daniel Tenbrinck, and Philipp Wacker for allowing me to use their material from earlier lectures!
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- Lecture: each Wednesday, 18:15 PM 19:45 PM German time, H11, videos from SS 2022 are available.
- Forum on StudOn to ask your questions, will be read regularly.



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- Forum on StudOn to ask your questions, will be read regularly.
- Exercises: Jan Krause and Ehsan Waiezi, take place online in the first week



• Exercises consist of both mathematical and implementation tasks



- Exercises consist of both mathematical and implementation tasks
- Students need to pass a written exam in presence at the end of the semester.
- Passing this course is mandatory for Master Data Science.



What is Data Science?

- relatively **new field** due to recent boost in digital transformation → digitization, IoT, paperless office, ...
- needs mathematics, computer science, and domain knowledge
- well-known examples for successfully harvesting and interpreting data:
 - \rightarrow Google
 - \rightarrow Spotify
 - \rightarrow Amazon

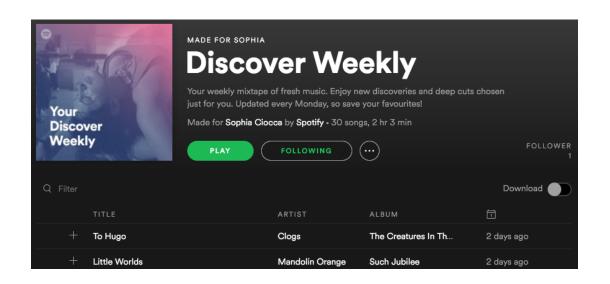






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Kids' Magic Secrets: Simple Magic Tricks **★★★☆☆** ▼ (8) \$9.95

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Data Science Cycle

iteratively...

- 1. extract and process / correct data
- 2. extract hypotheses
- 3. develop approaches, models, algorithms, implementations (from statistics, optimization, numerics and simulation, math theory, databases, AI,...)
- 4. use approaches to explore and understand data
- 5. derive predictions, decisions, consequences, recommendations for application domain
- 6. visualize data and results, possibly iterate

Your study programme covers these aspects, with specializations possible.

Many applications: logistics, mobility, health care, energy networks, ...



Data Science Cycle

Goals of this course

- learn fundamental math-based data science concepts and algorithms, know where to look it up
- understanding the underlying mathematical reasons why certain algorithms work well and others do not
- solve realistic problems



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Topics: unsupervised learning and supervised learning methods, e.g.

- clustering
- PCA, kernel methods, kernel-PCA
- statistical methods
- machine learning via neural networks
- graphical models,...



Some Data Science Examples

- prediction of solar injection in energy networks and best-possible curtailment decisions beforehand to avoid breakdown of the network
- learn customer / market preferences, produce accordingly and forecast price developments
- predict shelf life of food
- forecast development of chronic diseases, determine best possible medication and treatment
- forecast development of Covid-19, determine best possible reactions
- many more



Further Reading

- Hastie, Tibshirani, Friedman: The Elements of Statistical Learning: Data Mining, Inference, and Prediction, Springer
- www.deeplearningbook.org



Excursion: Some Background in Algorithms

This excursion is meant to give some brief and abstract introduction into algorithms. Further reading, e.g.:

- Thomas Cormen, Charles Leiserson, Ronald Rivest, and Cliff Stein: Introduction to Algorithms, MIT Press
- Thomas Ottmann and Peter Widmayer: Algorithmen und Datenstrukturen,
- Robert Sedgewick, Kevin Wayne: Algorithms, Addison-Wesley,
- Donald Knuth: The Art of Computer Programming, and many others.

For exemplary purposes, we consider a basic operation: sorting numbers.



Sorting is a basic operation in computer science.

- **Input:** n numbers $\langle a_1, a_2, \ldots, a_n \rangle$.
- Output: permutation $\langle a_1', a_2', \dots, a_n' \rangle$ such that $a_1' \leq a_2' \leq \dots \leq a_n'$.



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- instance: each concrete series of numbers, e.g.,
 ⟨32, 25, 13, 48, 39⟩ ⇒ ⟨13, 25, 32, 39, 48⟩
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- instance: each concrete series of numbers, e.g., $\langle 32, 25, 13, 48, 39 \rangle \Rightarrow \langle 13, 25, 32, 39, 48 \rangle$ (in general: all input that is necessary for determining a solution.)

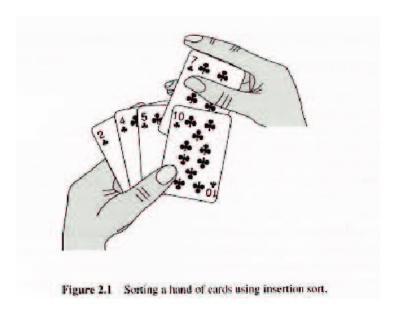
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An algorithm is called **correct**, if it **terminates** for all instances with a correct solution. It then **solves** the problem.



Insertion Sort



Aus: Cormen et al. (2001) "Algorithms", chpt. 2; MIT Press, Cambridge (MA)



Insertion Sort

As parameters, it has the array A and its length length(A). In the for-loop, the j-th element of the sequence is inserted in the correct position that is determined by the while-loop. In the latter we compare the element to be inserted (key) from 'right' to 'left' with each element from the sorted subsequence stored in A[0],...,A[j-1]. If key is smaller, it has to be insert further left. Therefore, we move A[i] one position to the right and decrease i by one in line 7. If the while-loop stops, key is inserted.

```
insertion_sort(A)
for j = 1 to (length(A)-1) do
  key = A[j]
  // insert A[j] into the sorted sequence A[1...j-1]
  i = j
  while (i > 0 and (A[i-1] > key) ) do
    A[i] = A[i-1]
    i = i-1
  end while
  A[i] = key
  end for
```



Insertion sort for the example sequence $\langle 5, 2, 4, 6, 1, 3 \rangle$

A[1]	A[2]	A[3]	A[4]	A[5]	A [6]
5	2	4	6	1	3
2	5	4	6	1	3
2	4	5	6	1	3
2	4	5	6	1	3
1	2	4	5	6	3
1	2	3	4	5	6



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- Termination: The while- and the for-loop always terminate.



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- time



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- 1 processor
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Question: Does the worst-case running time of insertion_sort grow linearly, quadratically, . . ., or even exponentially in n?



Running Time Insertion Sort

```
void insertion_sort(A){
  for j = 1 to < length(A-1) do
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    // insert A[j] into the sorted sequence A[1...j-1]
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For a sequence sorted in reverse order (worst case):

- The while-loop stops only when i = 0
- j = 1: 1 assignment A[i] = A[i 1]
- *j* = 2: 2 assignments
- . . .
- j = n 1: n 1 assignments



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```
• j = 1: 1 assignment A[i] = A[i - 1]
```

• . . .

•
$$j = n - 1$$
: $n - 1$ assignments

in total: $\sum_{i=1}^{n-1} i = \frac{n(n-1)}{2}$ many assignments; quadratically many



...more formally

Worst-Case Running Time

Idea: Consider only the characteristic behaviour as a function of the input size; ignore constants and terms of lower order.



...more formally

Worst-Case Running Time

Idea: Consider only the characteristic behaviour as a function of the input size; ignore constants and terms of lower order.

If the values of a function f are "less" than those of another function g for all n larger than some constant n_0 (up to a constant factor c), then asymptotically g is an upper bound for f, denoted by $f \in O(g)$.

More generally, the following symbols are often used for a function $g: \mathbb{N} \to \mathbb{R}$ (for upper bounds O, for lower bounds Ω and for equally fast growth Θ):

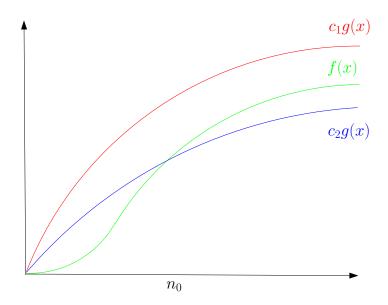
$$O(g) := \{f : \mathbb{N} \to \mathbb{R} \mid (\exists c, n_0 > 0) (\forall n \ge n_0) : 0 \le f(n) \le cg(n) \}$$

$$\Omega(g) := \{f \colon \mathbb{N} \to \mathbb{R} \mid (\exists c, n_0 > 0)(\forall n \geq n_0) : 0 \leq cg(n) \leq f(n)\}$$

$$\Theta(g) := \{f \colon \mathbb{N} \to \mathbb{R} \mid (\exists c_1, c_2, n_0 > 0) (\forall n \ge n_0) : c_1 g(n) \le f(n) \le c_2 g(n) \}$$



Worst-Case Running Time



• The worst case running time of insertion sort is $O(n^2)$.



Design of Algorithms

insertion sort: *incremental method.* different principle: *'divide and conquer'*

- divide problem in subproblems
- **conquer** the subproblems through recursive solution. (If small enough, solve them directly.)
- **combine** the solutions of the subproblems to a solution for the original problem.



Merge Sort

- **divide:** divide sequence of *n* numbers in the middle into two sub sequences.
- **conquer:** sort the subsequences recursively using *merge sort*.
- combine: merge the two subsequences to a sorted sequence.
- for sequences containing one element only nothing has to be done.



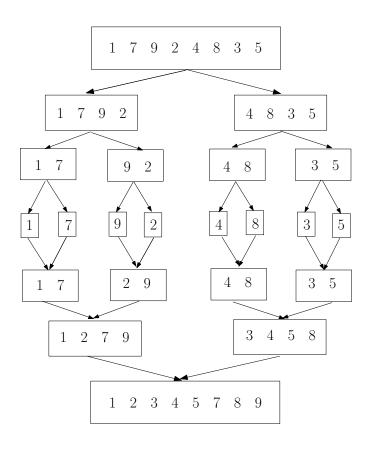
Merge sort

Sort the sequence stored in $A[p] \dots A[r]$:

```
void merge_sort(int[] A, int p, int r) {
  int q; /* Middle of the sequence */
  if (p < r) {    /* if p = r: only 1 element */
      q = p+((r-p)/2);
      merge_sort (A, p, q);    /* left subsequence */
      merge_sort (A, q+1, r);    /* right subsequence */
      merge (A, p, q, r);    /* merge subsequences */
  }
}</pre>
```



Illustration of Merge Sort





Merge Sort

It can be shown: worst-case running time of merge sort is $O(n \log n)$ (better than insertion sort)

In fact: any algorithm for sorting n numbers that uses only comparisons and moves of numbers needs at least $\Omega(n \log n)$.

BTW: try the shell-command **sort**!

This is the end of our excursion.

We will now return to the introduction of learning methods.