# Lab 02 – Complexity & friends

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Slides by Lorenzo De Carli

#### In this lab

- Make sure you have a working Python installation
- Refresh your knowledge of Python programming and syntax
- Make sure you are familiar with the tools you'll need in this class
- We'll discuss submission format at the end of this presentation

### Exercise 1 - Divide and conquer

- Memoization is an approach to algorithmic optimization where intermediate results are saved in a cache for future reuse
- A "cache" in this case is anywhere data can be accessed quickly, such as for example a Python array or a dictionary
- We are going to see how memoization can improve performance

#### Exercise 1/2

The exercise is going to be based on this code

```
def func(n):
   if n == 0 or n == 1:
     return n
   else:
     return func(n-1) + func(n-2)
```

#### Exercise 1/3

- First, answer these questions:
  - 1. What does this code do? [0.1 pts]
  - 2. Is this an example of a divide-and-conquer algorithm (think carefully about this one)? [0.1 pts]
  - 3. Give an expression for the time complexity of the algorithm [0.2 pts]

#### Exercise 1/4

- 4. Implement a version of the code which uses memoization to improve performance [0.2 pts]
- 5. Give an expression for the time complexity of the optimized algorithm [0.2 pts]
- 6. Time the original code and your improved version, for all integers between 0 and 35, and plot the results (output plots must be called ex1.6.1.jpg and ex1.6.2.jpg) [0.2 pts]

#### Exercise 1 – what to deliver

- Submit a Python file called ex1.py
- The file must contain:
  - The answer to question 1, 2 and 3 as code comments
  - The code implementing the answer to question 4 as a Python function
  - The answer to question 5 as code comments
  - Code implementing the timing requested by question 6
  - Do not submit the output plots

### Exercise 2 – interpolation search

For this exercise, we are going to use the following code

```
def interpolation_search(arr, x):
    low = 0
    high = len(arr) - 1
    while low <= high and x >= arr[low] and x <= arr[high]:
        pos = low + int(((float(high - low) / (arr[high] - arr[low])) * (x - arr[low])))
        if arr[pos] == x:
            return pos
        if arr[pos] < x:
            low = pos + 1
        else:
            high = pos - 1
        return -1</pre>
```

#### Exercise 2/1

- 1. Mention at least two aspects that make interpolation search better than binary search [0.1 pts]
- 2. Interpolation search assumes that data is uniformly distributed. What happens this data follows a different distribution? Will the performance be affected? Why? [0.2 pts]
- 3. If we wanted to modify interpolation search to follow a different distribution, which part of the code would be affected? [0.1 pts]

#### Exercise 2/2

- When comparing linear, binary and interpolation search:
  - 4. When is linear search your only option for searching data as binary and interpolation search may fail? [0.2 pts]
  - 5. In which case will linear search outperform both binary and interpolation search, and why? [0.2 pts]
  - 6. Is there a way to improve binary and interpolation search to solve this issue? [0.2 pts]

#### Exercise 2 – what to deliver

- Provide the answer to questions 1-6 as a markdown (.md) file called ex2.md
- Visual Studio Code already supports editing .md files

## Exercise 3 - profiling

- In class, we have discussed **benchmarking** as a way to measure code performance
- Now, let's learn about profiling

#### Exercise 3/2

#### We'll use this code for the exercise

```
import timeit
def sub_function(n):
  #sub function that calculates the factorial of n
  if n == 0:
    return 1
  else:
    return n * sub_function(n-1)
def test_function():
  data = []
  for i in range(10):
    data.append(sub_function(i))
  return data
def third_function():
# third function that calculates the square of the numbers from 0 to 999
  return [i**2 for i in range(100000000)]
test_function()
third_function()
```

#### Exercise 3/3

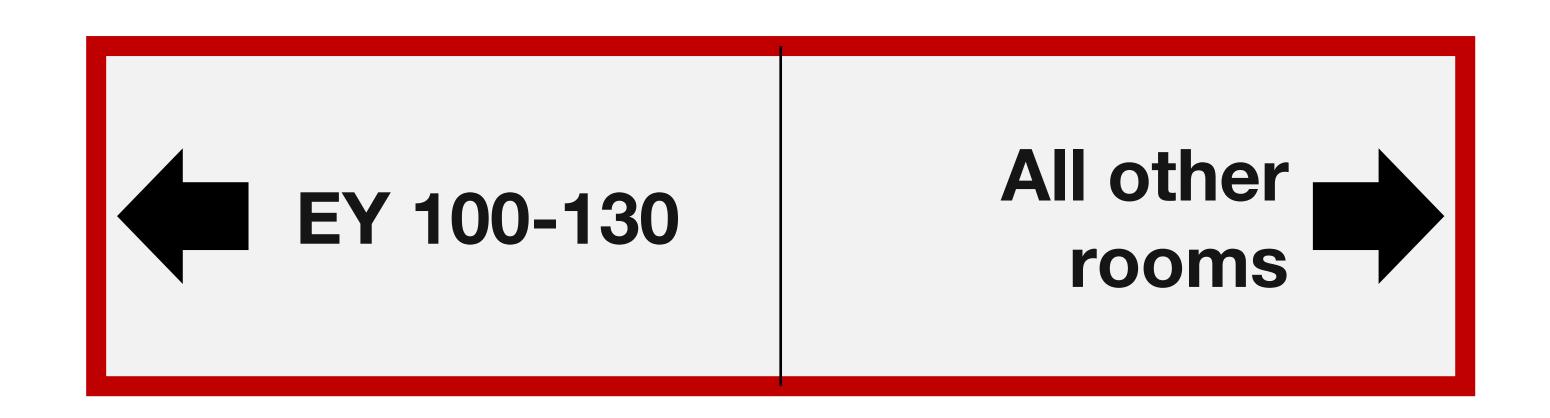
- First learn about the cProfile module in Python <a href="https://docs.python.org/3/library/profile.html">https://docs.python.org/3/library/profile.html</a>
- Then, answer the following questions:
  - 1. What is a profiler, and what does it do? [0.25 pts]
  - 2. How does profiling differs from benchmarking? [0.25 pts]
  - 3. Use a profiler to measure execution time of the program (skip function definitions) [0.25 pt]
  - 4. Discuss a sample output. Where does execution time go? [0.25 pts]

#### Exercise 3 – what to deliver

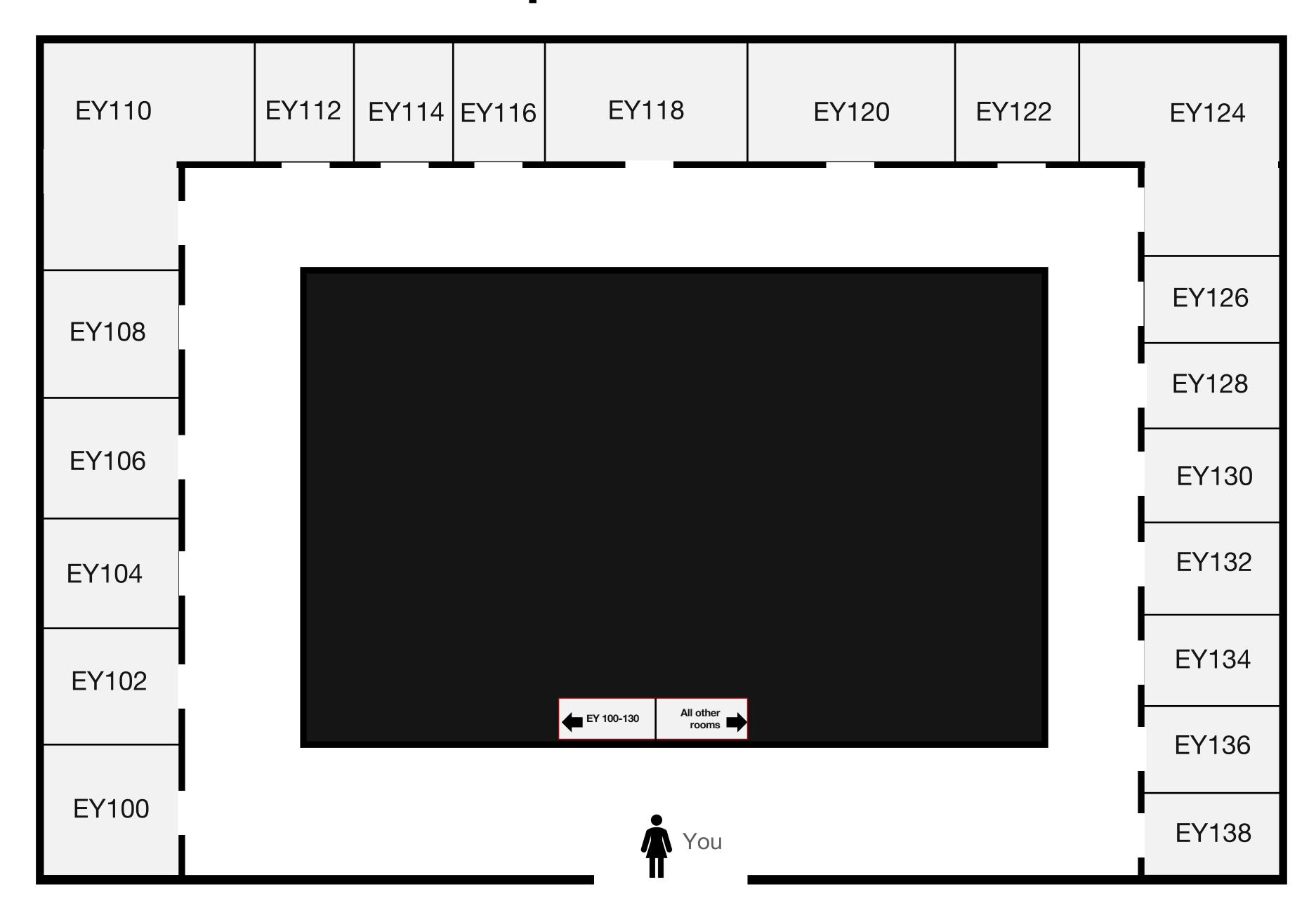
- Deliver a Python file named ex3.py
- The code should implement answers to question 3
- Answers to questions 1-4 should be provided in the comments to the same code

#### Exercise 4 – todo

- Imagine there is a new engineering building on Ucalgary campus ("Engineering Y"). You enter the building on the ground floor and you need to attend a lab in EY128.
- As you enter, you find the sign below:



### Exercise 4 /2 The floorplan of the first floor is described below



#### Exercise 4/3

- 1. Describe the algorithm you will use to find the room. Assume all the information you have is the one given by the sign; you have no knowledge of the floor plan [0.5 pts]
- 2. How many "steps" it will take to find room EY128? And what is a "step" in this case? [0.25 pts]
- 3. Is this a best-case scenario, worst-case scenario, or neither? [0.25 pts]
- 4. With this particular sign and floor layout, explain what a worst-case or best-case scenario would look like [0.5 pts]
- 5. Suppose after a few weeks in the term you memorize the layout of the floor. How would you improve the algorithm to make it more efficient? [0.5 pts]

#### Exercise 4 – what to deliver

• Provide the answer to questions 1-5 as a markdown (.md) file called ex4.md

### Exercise 5 – visualizing complexity

- In this exercise, you will implement basic search algorithms and attempts to confirm theoretical complexity findings with empirical measures
  - 1. Implement linear search and binary search [0.5 pts]
  - 2. Measure the performance of each on sorted vectors of 1000, 2000, 4000, 8000, 16000, 32000 elements. In each case, you must do the following for 1000 times, and compute the average [0.5 pts]:
    - a. Pick a random element in the vector
    - b. Measure the time it takes to find the element using timeit, using 100 iterations (number=100)

#### Exercise 5/2

- 3. Each plot should also interpolate the data points with an appropriate function. For example, linear complexity with a linear function, quadratic complexity with a quadratic function, etc. [0.5 pts]
  - You already know how to fit a linear function (from class). Fitting more complex functions can be accomplished using scipy.optimize.curve\_fit()
- 4. Discuss the results. For each interpolating function, describe (1) the type of function, and (2) the parameters of the function. Are the results what you expected? Why? [0.5 pts]

#### Exercise 5 – what to deliver

- Provide the answer to questions 1-3 as a Python file named ex5.py. The code file must include implementations of the algorithms, timing code, and plotting/interpolation code.
- Provide the answer to question 4 as comments within the same code.

#### What to deliver

- Upload a zip file to the "Lab 2" dropbox on D2L, containing the required content for every exercise:
  - ex1.py
  - ex2.md
  - ex3.py
  - ex4.md
  - ex5.py

### Grading rubric

- You get 3 pts for uploading a partial solution by end of lab
  - Must not be an empty file or irrelevant material
- Then, you'll have until 11:59PM of the day before the next lab to upload the complete solution.
   That will be graded as follows:
  - Exercise 1: 1 pts
  - Exercise 2: 1 pts
  - Exercise 3: 1 pts
  - Exercise 4: 2 pts
  - Exercise 5: 2 pts
  - Can upload the complete solution to the same dropbox

## That's all folks!