

# 11 Exercises

September 13, 2020

Import the modules that will be used.

```
[0]: import numpy as np
import matplotlib
import matplotlib.pyplot as plt

%matplotlib inline
```

## 0.1 Exercise 11.1

Determine by counting the number of mathematical operations the complexity of:

1. Dot product between two vectors
2. Matrix-vector product
3. Matrix-matrix product

for vectors of length  $n$  and matrices of size  $n \times n$ .

This is a reasoning exercise - you do not need to write a program. Express your answers in text and using LaTeX in a Markdown cell.

### 0.1.1 Optional

Test the complexity experimentally with your own functions for performing the operations, and with the NumPy ‘vectorised’ equivalents.

YOUR ANSWER HERE

1.  $O(1)$
2.  $O(n)$
3.  $O(n^2)$

## 0.2 Exercise 11.2

For the recursive factorial algorithm in Activity 04, determine the algorithmic complexity by inspecting your implementation of the algorithm. Test this against numerical experiments.

### 0.2.1 Solution

Recall the factorial algorithm from Activity 04.4:

```
[0]: def factorial(n):
      if n == 0:
          return 1
      else:
          return factorial(n - 1)*n
```

The function calls itself (recursively)  $n$  times, hence it has complexity  $O(n)$ . We test this below and plot the times.

```
[14]: # Create array of problem sizes we want to test (powers of 2)
      #N = np.arange(2, 8)
      # Create an array of random numbers
      #x = np.random.rand(N[-1])
      N = 100

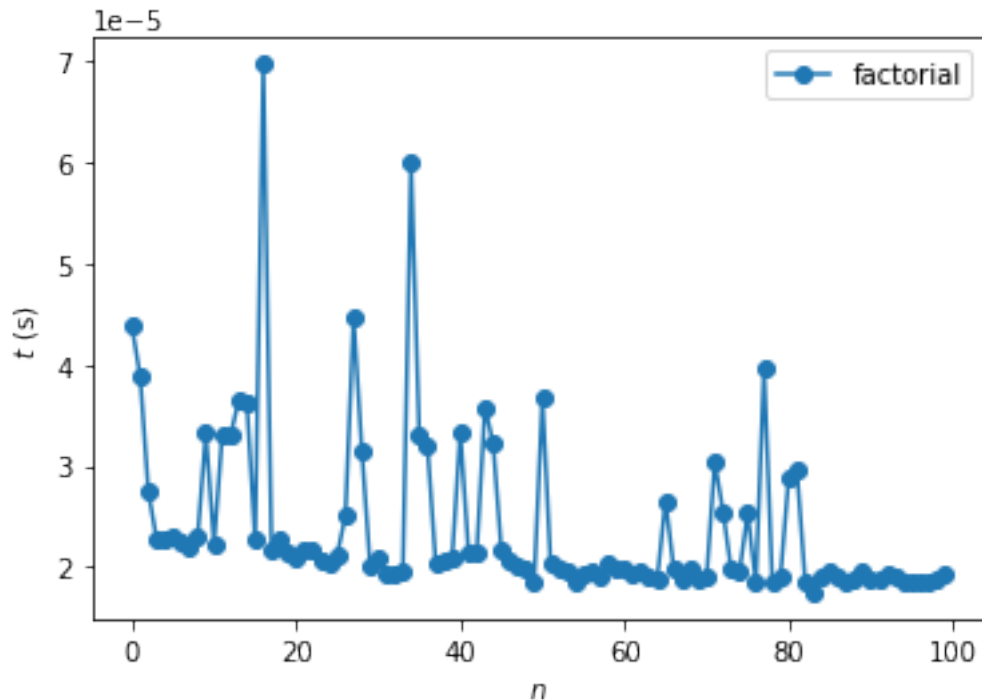
      # Time quicksort on arrays of different lengths
      times = []
      for n in range(N):
          t = %timeit -n1 -r1 -o -q factorial(N)
          times.append(t.best)

      # Plot quicksort timings
      plt.plot(times, marker='o', label='factorial')

      # Show reference line of  $O(n*\log(n))$ 
      #plt.loglog(N, 1e-6*N, label='$O(n)$')

      # Add labels
      plt.xlabel('$n$')
      plt.ylabel('$t$ (s)')
      plt.legend(loc=0)

      plt.show()
```



### 0.3 Exercise 11.3

Determine experimentally the complexity of computing the determinant of a matrix. You can generate an  $n \times n$  matrix using:

```
[0]: n = 100
      A = np.random.rand(n, n)
```

and the determinant can be computed by:

```
[0]: det = np.linalg.slogdet(A)
```

Be sure that you test for sufficiently large  $n$  to get into the ‘large’  $n$  regime.

#### 0.3.1 Solution

Time computation of determinant:

```
[0]: # Create array of problem sizes we want to test (powers of 2)
      N = 2**np.arange(2, 12)
      # Create an array of random numbers
      x = np.random.rand(N[-1])

      # Time quicksort on arrays of different lengths
      times = []
```

```
for n in N:
    t = %timeit -n1 -r1 -o -q det
    times.append(t.best)
```

Plot result:

```
[0]: # Plot quicksort timings
plt.loglog(N, times, marker='o', label='det')

# Show reference line of  $O(n \log(n))$ 
plt.loglog(N, 1e-6*N, label='$O(n)$')

# Add labels
plt.xlabel('$n$')
plt.ylabel('$t$ (s)')
plt.legend(loc=0)

plt.show()
```