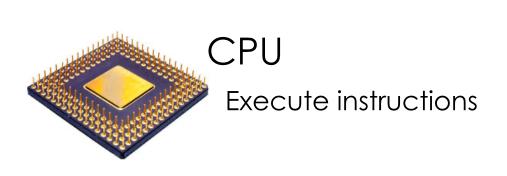
## What is programming?

- Writing instructions for computers to perform tasks

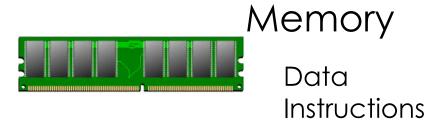


#### Instructions

- operators
- for loops
- if-else conditionals
- functions

• • •

Classes



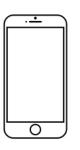
#### Data

- lists
- tuples
- dictionaries
- sets

Scripts
Modules,
Packages,
Libraries

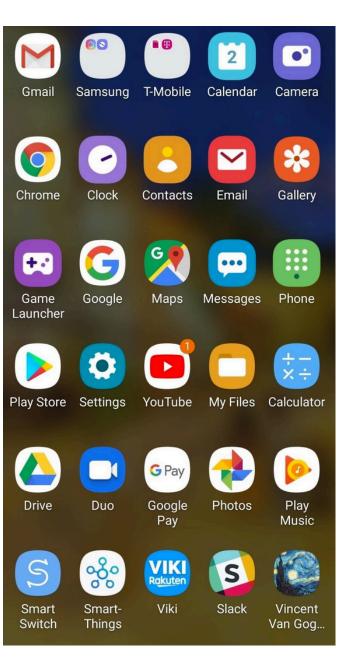
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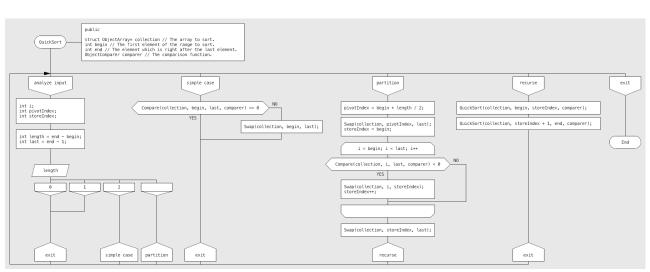






```
79
80
81
        <del>%%%%%%%%%%%%%%%%%%%%%%%%%%</del>
82
        % pos input
83
84 -
        d = 2*pi/n input azimuth;
85
86
        mamp = 0.35;
87
88
      if use_2D_input && n_input_elevation>2
89
            mamp = mamp*0.7;
90 -
91
92 -
      if contains(rule_id_list{rule_id}, 'Pre-synaptic')
93 -
            mamp = mamp/3;
94 -
95
96
        disp(['max amp of input=' num2str(mamp)]);
97
98
        %mc = xpos_radian;
99
100
        mcshift = ceil(mod(xpos_radian,2*pi)/(2*pi)*size(sim_visual_neurons,1));
101
102 -
        testing if input is strong enough = 0;
103 -
      if testing_if_input_is_strong_enough
104 -
            id1 = ceil(numel(xpos_radian)/400*110);
105 -
            id2 = ceil(numel(xpos_radian)/400*350);
106
            mcshift(id1:id2) = mcshift(id1:id2) + ceil(size(sim_visual_neurons,1)/2);
107 -
        end
108
109
```

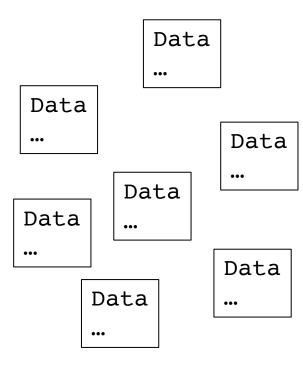




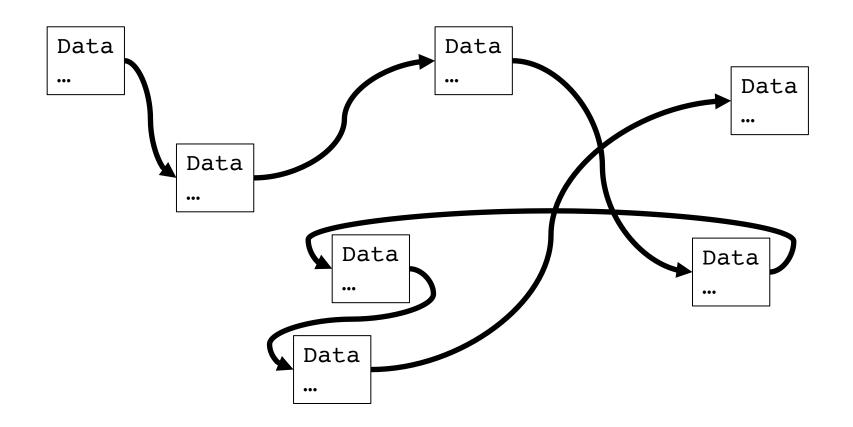
JupyterLab Reference X Syn\_counter.ipynb Python 3 [4]: import pandas as pd # to read in the synapse data import numpy as np import matplotlib.pyplot as plt # for plotting # for OS specific operations such as directory/file manipulations import os # The directory of this file. Make sure you have a "figures" sub-directory in this directory. os.chdir('/Users/kims11/Documents/SS\_research/codes/projects/synapses') syn\_path = os.getcwd() [6]: # V0 is based on the segmentation with the following timestamp: # 2020-03-29T06:32:03.479060 # datetime.datetime(2020, 3, 29, 6, 32, 3, 479060) # Timezone in our system is always UTC. # One can pin a timezone in neuroglancer to view the segmentation at point in time. #flywire\_buhman\_wiring\_v0.csv—2020-03-2796/32:03.479060 —datetime.datetime(2020, 3, 29, 6, 32, 3, 479060) #flywire\_buhman\_wiring\_v1.csv—2020-05-02706:26:05.700341 —datetime.datetime(2020, 5, 2, 6, 26, 5, 700341) #flywire\_buhman\_wiring\_v2.csv—2020-05-31733:39:05.100902—datetime.datetime(2020, 5, 13, 23, 39, 5, 100902) # Put the .csv data in the "syn\_path" directory defined above. Then continue running code. #syn\_fn = os.path.join(syn\_path, 'flywire\_buhman\_wiring\_v0.csv'); # old version syn\_fn = os.path.join(syn\_path, 'flywire\_buhman\_wiring\_v3.csv'); print(syn\_fn) # Read the data #syn\_df = pd.read\_csv(syn\_fn)#, dtype="uint64") # Use "uint64" to preserve the precision of the synapse data. This reading process should take 5 minutes or l syn\_df = pd.read\_csv(syn\_fn), dtype=np.uint64) # Use "uint64" to preserve the precision of the synapse data. This reading process should take 5 minutes or le /Users/kims11/Documents/SS\_research/codes/projects/synapses/flywire\_buhman\_wiring\_v3.csv [7]: # A basic Panda dataframe command syn\_df.head() Unnamed: 0 pre pt root id pre pt sv id pre\_pt\_position ctr\_pt\_position post pt root id post pt sv id 0 720575940620535651 76916748439627068 [366196 244760 73920] [366146 244770 73900] 720575940631123034 76916748439628340 [366296 244740 73960] 1 720575940620535651 76916748439635341 [364864 245640 73960] [364942 245666 73960] 720575940616174657 76916748439614229 [364708 245588 73960]

2 720575940616174657 76916748439614229 [364564 244904 73960] [364506 244936 73940] 720575940545890960 76916748439623112 [364680 244840 74000] 3 720575940620535651 76916748439632311 [364600 244468 74200] [364624 244530 74200] 720575940627209594 76916748439633329 [364552 244344 74200] 3 4 720575940636149614 76916748439652476 [364952 244764 74240] [364998 244748 74260] 720575940620535651 76916748439632348 [364860 244796 74200]

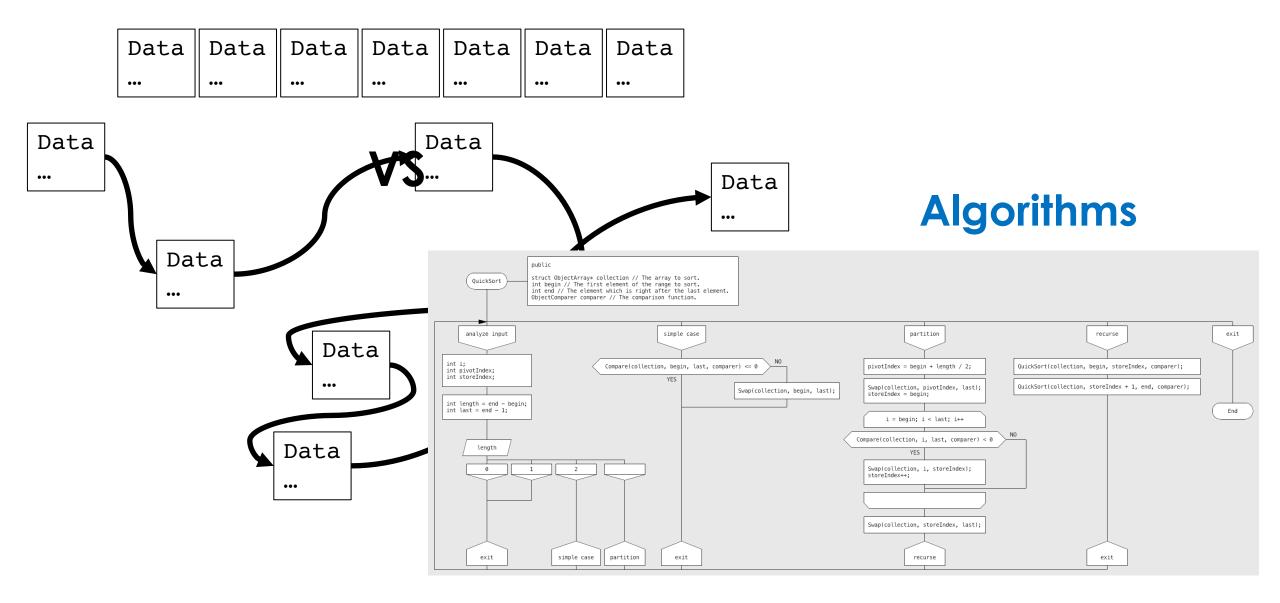
1: # A basic Panda dataframe command syn\_df.tail()



| Data |  |
|------|------|------|------|------|------|------|--|



#### **Data Structures**





Data Data Data **Linked List** Data Data Data Data •••



Fast for search Slow for resizing

#### **Linked List**



Slow for search Fast for resizing

How fast is the algorithm?



# Big O

**Big O** notation, also called **Landau's symbol**, describes how running time (or memory space) for your algorithm grow as the size of the input to the algorithm grows

**Big O** notation describes the **asymptotic behavior** of the algorithm

| Data |
|------|------|------|------|------|------|------|
| •••  | •••  | •••  | •••  | •••  | •••  | •••  |

#### **Linked List**



```
def add_a_number_to_all_data(data, number):
    for a in data:
        a += number

(This is not real code. It is conceptual logical flow of a hypothetical function.)
```

as  $N \to \infty$   $time \sim N$ 

 $time = a \cdot N + b$ 

O(N) (order of N)

| Data |
|------|------|------|------|------|------|------|
| •••  | •••  | •••  | •••  | •••  | •••  | •••  |

#### **Linked List**



```
def add_all_pairs_of_data(data):
    matrix = matrix_init()
    for i, a in enumerate(data):
        for j, b in enumerate(data):
            matrix(i,j) += a+b
    return matrix

(This is not real code. It is conceptual logical flow of a hypothetical function.)
```

$$time = a \cdot N^2 + b$$
as  $N \to \infty$ 

$$time \sim N^2$$
 $O(N^2)$  (order of N squared)

| Data |
|------|------|------|------|------|------|------|
| •••  | •••  | •••  | •••  | •••  | •••  | •••  |

#### **Linked List**



```
def return_the_value_at_position_index_n_in_array(data, n): return data[n] O(1) \quad (\text{order of 1: constant time})
```

There are common patterns in programming.

Two major categories of patterns: Data structures and Algorithms

The choice of data structure affect the implementation of algorithms and its process time.

### **Array**

| Data |
|------|------|------|------|------|------|------|
| •••  | •••  | •••  | •••  | •••  | •••  | •••  |

Big O

#### **Linked List**

