Techniques

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1 Some python knowledge, and techniques useful for your assignment

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Table of Contents

- 1 Introduction
- 2 Strings (unchangable series of characters)
- 3 Tuples (unchangeble lists)
- 4 Lists are just like tuples, but can be changed in place
- 5 (numpy) arrays, so-called ndarrays (multi-dimensional arrays)
- 6 Some ways to generate arrays
- 7 For-loops with zipping and enumerate
- 8 Superposition, using logical indexing to handle switch times
- 9 Some elegant logical indexing, which is extremely useful
- 10 List comprehensions to generate and filter lists
- 11 Conclusions

1.1 Introduction

To help you get up and going with your assignment, I explain in this notebook some features of Python and some techniques in Python that will prove useful.

1.2 Strings (unchangable series of characters)

The function dir(a) gives all objects associated with a, a string. Show only the public ones, i.e. the ones that do not start with '_'.

Here are the functions that can be applied on strings. But there is none with which you can change an existing string; you can, however, replace your string it with a new one.

```
In [3]: # I use a list comprehension to generate a list from an iterable (see further down
                           # for explanation)
                          methods = [x for x in dir(a) if not x.startswith('_')]
                          print(methods)
['capitalize', 'casefold', 'center', 'count', 'encode', 'endswith', 'expandtabs', 'find', 'formation of the count', 'encode', 'endswith', 
         Now apply some of these so-called string methods on the string a.
In [4]: print(a.title(), ", result is a with capitalized first letter.")
                           # Note you get a new string, a is still the same
                          print(a, ", a is still unchanged.")
This Is A String , result is a with capitalized first letter.
This is a string , a is still unchanged.
In [5]: a.upper()
Out[5]: 'THIS IS A STRING'
In [6]: a.split() # split the string on whitespace
Out[6]: ['This', 'is', 'a', 'string']
In [7]: a.startswith('John')
Out[7]: False
In [8]: a.endswith('ing')
Out[8]: True
In [9]: a.upper().endswith('ING')
Out[9]: True
```

Insert values into a string on successive locations indicated by {}. This is done by applying the function format on the string. The aguments of format(..) are the values that will be formatted and placed in the successive placeholders {}.

We can put code inside these placeholders to specify how each value should be formatted. Read the documentation to find out. There is a wealth of possibilities.

```
In [10]: 'Today is {} {} in the year {}.'.format('Monday', 3, 'Feb', 2019)
Out[10]: 'Today is Monday 3 Feb in the year 2019.'
```

1.3 Tuples (unchangeble lists)

Selection of more than one item at once to get a sub-tuple, is done by by slicing like so: a[:] = 'all' of tuple a, a[:5] means a from start up to but not including item 5, a[4:] means a from item 4 to the last item.

Examples

By slicing (with the colon:) you always get a tuple back, unless the result is only a single item, then you get the item itself a[0:1] would yield the first item (0 up to but not including 1)])

A tuple consisting of only one item is indicated with a comma:

```
(2,) b is a tuple, due to the comma
2 c is just aa number, not a tuple
```

A series of items separated by comma's is also a tuple, like in the arguments of a function.

```
In [15]: 1, 2, 3, 4
Out[15]: (1, 2, 3, 4)
```

A single number or any single object ended with a comma indicates a tuple of one item

1.4 Lists are just like tuples, but can be changed in place

Lists are demarcated by straight brackets [] instead of parentheses (). But they work further exactly like tuples:

Don't use a comma to indicate a list a list of only one item.

[3] means a list of one item (item = 3) [3,] means turn the tuple (3,) into a list. It is namely the same as [(3,)] because 3, is same as (3,) as was explained above.

A list is changable in place. So you can insert items and extend it or sort it, all in place, which you cannot with a tuple, which you have to replace.

Note we have appended an entire list to the list a and we did so in-place, meaning that a itself has now changed.

Notice that a now consists of 4 items. First three strings and then an entire list as the fourth item, which itself also contains items and even a list. But a has only four items after the append().

This is what you can do with a list (dir(a) gives the methods associated with list a, so that the can be applied on a with the dot ., like a.count(), a.append(..) etc.):

Only count and index, no append insert or pop or sort, i.e. anything that could alter the tuple. Hence, trying to append something tuple b won't work. But we can replace b.

```
In [22]: #('one', 'two').append('cow') # this gives an error, because a tuple has no attribute
['one', 'two'].append('three') # works
```

However, because append changes the list in place, it does not produce output so we have no result unless we assign the list to a variable.

None, This is "b", it is a "None" because a.append() changes "a" in place and yields no output. ['one', 'two', 'cow'], This is "a", it was changed in place.

1.5 (numpy) arrays, so-called ndarrays (multi-dimensional arrays)

Arrays may look like lists, but they are not. In a lists each item can be of a different type, as was shown above, but in an array all items must be of the same type like floating point numbers. An array forms one contiguous block in memory, in which each item occupies the same number of bytes (a floating point number occupies 8 bytes, an integer 4 bytes). This means that items can be accessed at full speed using indexing and slicing. You can also add arrays of the same size together, take the power on a whole array or other functions etc. You cannot do this with a list, because items in the list may be of completely different types. In fact, the items in the list are stored on different locations in memory, while the list is just a number of pointers that indicate these places in memory so that they can be accessed indirectly, yet still quite fast.

If you multiply a list, you copy the list and append it like so:

But if you multiply an array, then each item in the array is multiplied as expected. First generate an array (a numpy ndarray):

```
In [25]: b = np.array([3, 4, 5, 2, 9, 8])
```

b is not a list, but an array, because np.array(...) function turns the given list into an array.

```
5 \text{ times } b = [15 \ 20 \ 25 \ 10 \ 45 \ 40]
In [28]: print('b + b + b = ', b + b + b)
b + b + b = [9 12 15 6 27 24]
In [29]: print('b * 2 * b**3 = ', b * 2 * b ** 3)
b * 2 * b**3 = [ 162 512 1250
                                       32 13122 8192]
In [30]: print(np.sqrt(b))
Γ1.73205081 2.
                       2.23606798 1.41421356 3.
                                                         2.82842712]
In [31]: print(np.sin(b * np.exp(-5)))
[0.02021246 0.02694853 0.03368336 0.01347549 0.06060436 0.05387748]
1.6 Some ways to generate arrays
In [32]: a = np.array([3, 2, 4, 5])
         print(a)
[3 2 4 5]
In [33]: a = np.arange(3, 7, 0.3) \# start at 3 upto (not including 7, with steps of 0.3)
         print(a)
[3. 3.3 3.6 3.9 4.2 4.5 4.8 5.1 5.4 5.7 6. 6.3 6.6 6.9]
In [34]: a = np.linspace(1, 4, 11) # give 11 points starting at 1 and ending at 4
         print(a)
[1. 1.3 1.6 1.9 2.2 2.5 2.8 3.1 3.4 3.7 4.]
In [35]: a = np.logspace(-2, 1, 31) # 10**(-2) to 10**1 in 30 steps (31 values)
         print(a)
[ 0.01
              0.01258925  0.01584893  0.01995262  0.02511886  0.03162278
  0.03981072 0.05011872 0.06309573 0.07943282 0.1
                                                                0.12589254
  0.15848932 \quad 0.19952623 \quad 0.25118864 \quad 0.31622777 \quad 0.39810717 \quad 0.50118723
                                       1.25892541 1.58489319 1.99526231
 0.63095734 0.79432823 1.
 2.51188643 3.16227766 3.98107171 5.01187234 6.30957344 7.94328235
 10.
            1
```

```
In [36]: a = np.zeros_like(a)
       print(a)
0. 0. 0. 0. 0. 0. 0.]
In [37]: a = 5 * np.ones((3, 2)) # 2D array with 3 rows and 2 columns
       print(a)
[[5. 5.]
[5. 5.]
[5. 5.]]
In [38]: a[:, 1] = np.pi # replace all rows of column 2 (with index 1) by np.pi
       print(a)
ſſ5.
           3.141592657
[5.
           3.14159265]
[5.
           3.14159265]]
In [39]: b = np.sin(2 * np.pi * a) # apply function on all values in array
       print(b)
[[-1.2246468e-15 7.7685322e-01]
[-1.2246468e-15 7.7685322e-01]
[-1.2246468e-15 7.7685322e-01]]
```

1.7 For-loops with zipping and enumerate

A for loop cycles over an iterable, which is something you can iterate over like a list, a tuple, an array, a string. Here are some examples.

In each new cycle of the loop, the next item of the iterble is taken and can be used inside the loop. But very often we have a set of iterable like the array of x-coordinates of the wells, the array of y-coordinates of the wells and the list of extractions from the wells. In each cycle of the loop we need the next x, the next y and the next Q at the same time. To deal with such a set of iterables in a loop just put them in a zip(...) function.

Oftentimes you not only want the next times from the iterable but also have a counter, an index while looping. This counter can be obtained by packing the iterable or the entire zip(..) with all its iterables in an enumerate(..) function. Note that counting starts with 0 in python, not 1.

This type of looping is efficient and extremely useful in practice. For instance we have 5 wells so 5 x vales, 5 y values 5 flows for which we want the drawdown s at a given point x0, y0

```
In [43]: kD = 600 # m2/d
    S = 0.1
    t = 1.3 # d
    xWells = [34, 123, 45. -70, -80.]
    yWells = [-23., 45., -50., -30., 20.]
    Qwells = [1200, 600., 350., 1200., 300]

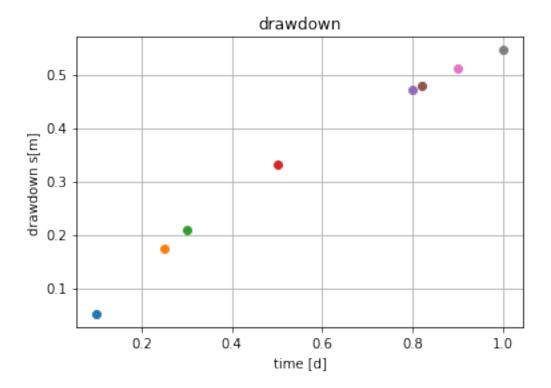
x0, y0 = 0.3, 6.7 # multiple assignment, handy

times = [0.1, 0.25, 0.3, 0.5, 0.8, 0.82, 0.9, 1.0] # days

# plot individual times as a dot
    plt.title('drawdown')
    plt.xlabel('time [d]')
    plt.ylabel('drawdown s[m]')
    plt.grid()

for t in times:
```

```
s = 0
for xw, yw, Qw in zip(xWells, yWells, Qwells):
    r = np.sqrt((x0 - xw)**2 + (y0 - yw)**2)
    s += Qw / (4 * np.pi * kD) * exp1(r**2 * S / (4 * kD * t))
    plt.plot(t, s, 'o') # plot the point as a dot, color is automatic
plt.show()
```



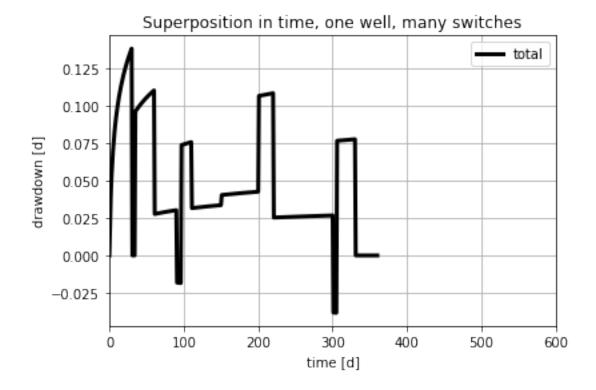
1.8 Superposition, using logical indexing to handle switch times

When dealing with superposition, we have a set of so called switch times at which the boundary conditions change and we have an array of times with values at small intervals to obtain sufficient values to produce a nice, detailed graph.

During the superposition loop, we need to obtain the results of for instance the head change s at times larger than the the switch time, e.g. st3 when the well was swithed on. We compute its results for times t - st3, but only for those times for which t > st3.

We can select those times by logical indexing. That is, if t is an array of times and ts3 is a single time, then t > ts3 is an array of booleans, an array of True/False values of the same size as the array t, such that the elements that are True correspond with the condition t > ts3. We can then address the correct values of an array s, which has the same size as the array t. We do this like so s[t > ts3]. Or, in two steps: I = t > ts3 and then select the right values like so s[I].

```
= np.array([300, 0, 200, 50, -30, 120, 50, 60, 150, 35, -50, 100, 0])
         # changes of flow. hstack((a, b)) glues two arrays or a value and array together
         dQ = np.hstack((Q[0], np.diff(Q)))
         \#print('dQ = ', dQ)
In [45]: xw, yw, x0, y0 = 100, 50, 0.5, -3 # assigning all at once, well and obs point coordinate
        r = np.sqrt((xw - x0)**2 + (yw - y0)**2)
         # set up the plot
        plt.title('Superposition in time, one well, many switches')
        plt.xlabel('time [d]')
        plt.ylabel('drawdown [d]')
        plt.xlim((0, 600))
        plt.grid()
         # Initialize the head changes to an array of all zeroes
         s = np.zeros_like(times)
         # Do the superposition
         for st, Q in zip(swt, dQ):
             I = times > st # logical array I telling which times > st
             # print(I) # shows the booleans
             # effect of this well only, note times[I] are used onlyu
             ds = Q/(4 * np.pi * kD) * exp1(r**2 * S /(4 * kD * times[I] - st))
             \#plt.plot(times[I], ds, label='st = {:.0f} d'.format(st))
             s[I] += ds # add them to s pertaining to the right times[I] !!
         plt.plot(times, s, 'k', lw=3, label='total')
         plt.legend()
         plt.show()
```



It's extremely useful to get used to logical indexing of arrays. A logical index is an array of booleans (i.e. True/False values) of the same shape as the target arrays that can be used as indices. Like so

1.9 Some elegant logical indexing, which is extremely useful

[False True False True True True False True]

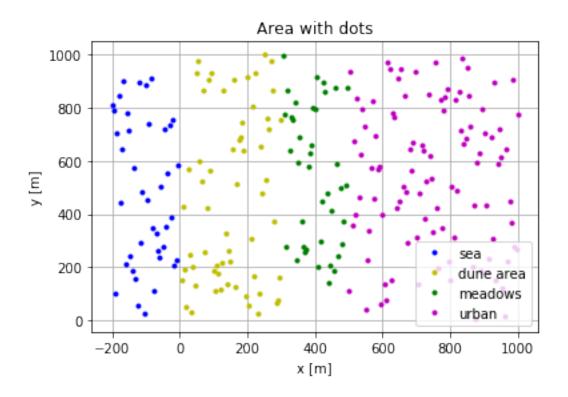
Then use the boolean array to pick values

We applied logical indexing above to select the times that were greater than ts, and did something with those times only.

Note that while the boolean array is as along as the target array. So boolean array I is as long as floating point array a, a[I] is smaller, i.e. it's as long as the number of True values in I

Here's an example of elegant usage of logical indexing to select subareas.

```
In [49]: x = np.linspace(-200, 1000, 241) #; print(x)
        y = np.random.rand(len(x)) * 1000 # random number as many as there are x-values
        # logical indexing, specifying subarea, names are intuitive
        sea = x < 0
        dunes = np.logical_and(x > 0, x <=300)
        grass = np.logical_and(x > 300, x < 500)
        urban = x >= 500
        plt.title('Area with dots')
        plt.xlabel('x [m]')
        plt.ylabel('y [m]')
        plt.grid()
        plt.plot(x[sea], y[sea], 'b.', label='sea') # b. means blue as dots
        plt.plot(x[dunes], y[dunes], 'y.', label='dune area') # y. means yellow as dots
        plt.plot(x[grass], y[grass], 'g.', label = 'meadows') # g. means green as dots
        plt.plot(x[urban], y[urban], 'm.', label= 'urban') # m. means magenta as dots
        plt.legend()
        plt.show()
```



1.10 List comprehensions to generate and filter lists

Start with an iterable, here an array, but it could by any iterable like a tuple or an list.

In
$$[50]$$
: a = np.array($[-3, 2, -1, 1.2, 3.2, 0.7, 3.2, -3, 5.1]$)

By way of trivial start, generate a list from the items in this iterable

In [51]:
$$b = [y \text{ for } y \text{ in } a] \# most \ basic \ list \ comprehension, \ b \ will \ be \ the \ same \ as \ a \ print('b = ', b)$$

$$b = [-3.0, 2.0, -1.0, 1.2, 3.2, 0.7, 3.2, -3.0, 5.1]$$

Notice that b is not an array, it is now a list. Use np.array(b) to make it an array if you need it.

Next, select only the items from the array a that are larger than some value. This is called filtering the iterable:

In [52]:
$$b = [x \text{ for } x \text{ in a if } x > 1.1] \# same but filtering out some values according to condition print(' $b = '$, b)$$

$$b = [2.0, 1.2, 3.2, 3.2, 5.1]$$

And of course, you got a list.

Now while you get the items from the list one by one in the comprehension, so something with them before putting them in the list. Here we take the logarithm of each item and we only take the items that fulfill the if-criterion of the comprehension.

```
[1.3862943611198906, 0.36464311358790924, 2.3263016196113617, -0.7133498878774649, 2.32630161961
```

From this example it becomes clear that comprehensions can be really advanced and are very flexible.

As another example, we take only the times of type str and we'll convert these strings into their uppercase variants before putting them in the list:

Clearly, comprehensions are extremely flexible for generating new lists from iterables (an iterable can be a list, tuple, array, string, i.e. is somethign you can iterate over like in for this in that. Then that is the iterable and this is an item from the iterable.

You can always turn the obtained list into an array if it contains only numbers:

```
In [55]: a = np.random.rand(10) # 10 random numbers
         print(a)
[0.35336559 0.37195635 0.27155251 0.0531265 0.99750573 0.41746925
0.95680672 0.72585634 0.35469918 0.64734761]
In [56]: b = [y \text{ for } y \text{ in a if } y > 0.2]
         print('b = ', b) # list
         print()
         c = np.array([y for y in a if y > 0.2])
         print('c = ', c) # an array
         print()
         d = ['y = {:.2f}'.format(y) for y in a if y < 0.5] # an array of strings
         print('d = ', d)
         print('joined into a single string: ', '; '.join(d)) # this is now a string
b = [0.35336559212459806, 0.37195634535238964, 0.27155250883278936, 0.997505734275096, 0.417469]
c = [0.35336559 \ 0.37195635 \ 0.27155251 \ 0.99750573 \ 0.41746925 \ 0.95680672
0.72585634 0.35469918 0.64734761]
d = ['y = 0.35', 'y = 0.37', 'y = 0.27', 'y = 0.05', 'y = 0.42', 'y = 0.35']
```

joined into a single string: y = 0.35; y = 0.37; y = 0.27; y = 0.05; y = 0.42; y = 0.35

1.11 Conclusions

- We showed strings, tuples, lists and numerical arrays (numpy ndarrays), which are all iterables (can be iterated over)
- We showed some basic logical indexing
- And used in loops
- We looped using several iterable in parallel to get values of them that belong together like the x, y and Q of successive wells.
- We have only used one-dimensional arrays here. That is we only scratched the surface of the possibilities.
- We did not use list of lists or list of tuples or tuples of lists etc. either.
- We only used basic examples of list-comprehensions.
- We did not look at dicts and dict comprehensions (generating dictionaries)
- We only showed the simplist logical indexing
- There is still a world of plotting possibilities to be explored (matplotlib gallery)
- But it's a good start.