

# Iterables vs. Iterators vs. Generators

Let's explore the differences between...

- a container
- an iterable
- an iterator
- a generator
- a generator expression
- a {list, set, dict} comprehension



## Containers

- data structures which hold elements
- support membership tests
- live in memory
- typically hold all their values in memory
- e.g., string, list, tuple, set, dict
- an object is a container when it can be asked whether it *contains* a certain element

```
In [1]: 1 in [1, 2, 3], 0 in [1, 2, 3]
```

```
Out[1]: (True, False)
```

```
In [2]: 4 in {4, 5, 6}, 1 in {4, 5, 6}
```

```
Out[2]: (True, False)
```

```
In [3]: 44 in ('Obama', 'Barack', 44, 2008, 'left')
```

```
True
```

```
Out[3]:
```

```
In [4]:
```

```
# for dicts, membership checks the keys, not the values
44 in { 43: 'Bush', 44: 'Obama'}, 'Bush' in { 43: 'Bush', 44: 'Obama' }
```

```
Out[4]: (True, False)
```

```
In [5]:
```

```
'J' in 'Steve Jobs', 'Job' in 'Steve Jobs', 'Jobs' not in 'Carlos Jobim'
```

```
Out[5]: (True, True, True)
```

## Iterables

- any object, not necessarily a data structure, that can return an iterator (with the purpose of returning all of its elements)
- the `__iter__()` function returns an iterator
  - ...therefore, any object which has the `__iter__()` method is an iterable
- most containers are also iterable
- many more things are iterable as well (e.g., open files, open sockets, etc.)

```
In [6]:
```

```
mylist = [1, 2, 3]
iter1 = iter(mylist)
print(type(iter1))
iter2 = mylist.__iter__() # iter() maps to __iter__()
print(type(iter2))
next(iter1)
```

```
<class 'list_iterator'>
<class 'list_iterator'>
```

```
Out[6]: 1
```

```
In [7]:
```

```
iter1.__next__() # next() maps to __next__()
#next(iter1)
```

```
Out[7]: 2
```

```
In [8]:
```

```
next(iter1)
```

```
Out[8]: 3
```

```
In [9]: next(iter1)
```

```
-----  
StopIteration                                Traceback (most recent call last)  
/tmp/ipykernel_26495/3389250325.py in <module>  
----> 1 next(iter1)
```

```
StopIteration:
```

```
In [ ]: next(iter2)
```

```
In [ ]: type(mylist), type(iter1), type(iter2)
```

```
In [ ]: # a list is iterable, but it is not its own iterator  
next(mylist)
```

```
In [ ]: iter(mylist) is mylist
```

```
In [ ]: mylistiter = iter(mylist)  
print('%x %x' % (id(mylist), id(mylistiter)))
```

```
In [ ]: iter(mylistiter) is mylistiter
```

## When we write...

```
mylist = [1, 2, 3]  
for x in mylist:  
    ...
```

...this is what happens



We can see this by disassembling the Python code...

In [ ]:

```
import dis
mylist = [1, 2, 3]
total = 0
dis.dis('for item in mylist: total += item')
```

## So what is an iterator?

- a stateful object that produces the next value when you call `next()` on it
- any object that has a `__next__()` method is therefore an iterator
- how it produces a value is irrelevant
- in other words, an iterator is a value factory
  - each time you ask it for "the next" value, it knows how to compute it because it holds internal state

In [ ]:

```
# Let's see how an iterator works...
mylist = [13, 46, -3, 'Go!']
myiter = iter(mylist) # get the list iterator

try:
    while True:
        val = next(myiter)
        print(val, end=' ')
except StopIteration:
    print('Stop!')
```

## Let's build our own iterator!

In [ ]:

```
class Fibonacci(object):
    def __init__(self):
        self.prev = 0
        self.curr = 1

    def __iter__(self):
```

```

    return self

def __next__(self):
    """
    each call to next() does two important things:

    1. modify its state for the subsequent next() call
    2. produces a result for the current call
    """
    value = self.curr
    self.curr += self.prev
    self.prev = value
    if value > 1000:
        raise StopIteration
    return value

# Note that this class is both an iterable due to __iter__()
# method and its own iterator, due to __next__() method!

f = Fibonacci()
print(next(f), next(f), 'before the for loop')

for num in f:
    print(num, end=' ')

```

## Lab: Iterators

Write your own iterator class which takes an iterable and each time it's invoked, it returns a *random* element. The iterator should stop (i.e., `raise` the `StopIteration` exception) when it has returned all elements of the iterable.

Example: `MyRandomIterator([1, 2, 3])` might return

```

2
3
1
...then raise StopIteration

```

In [ ]:

```

from collections import namedtuple

card = namedtuple('Card', ['rank', 'suit'])

```

```

class DeckOfCards:
    ranks = list(range(2, 11)) + list('JQKA')
    suits = 'clubs diamonds hearts spades'.split()

    def __init__(self):
        self._cards = [card(rank, suit) for suit in self.suits for rank in self.ranks]

    def __str__(self):
        time_to_print = [str(x.rank) + ' of ' + str(x.suit) for x in self._cards]
        return str('\n'.join(time_to_print))

    # use the iterator from the underlying list
    # instead of relying on the "default" Python iterator
    # which makes use of __getitem__ and __len__ (if it exists)
    # (__getitem__ should be enough)
    def __iter__(self):
        return iter(self._cards)

deck = DeckOfCards()
for card in deck:
    print(card)

```

## Generators

- a generator allows you to write iterators much like the Fibonacci iterator above but in an elegant, succinct syntax that avoids writing classes with `__iter__()` and `__next__()` methods
- every generator is an iterator (but not vice versa!)
- a generator is a factory that lazily produces values
- two types: generator *functions* and generator *expressions*

## The `yield` statement

- before we jump into generators, let's take an in-depth look at what makes them possible...
- when a normal Python function is invoked, execution starts at the first line and continues until a `return` statement is encountered or an exception is thrown (remember that "falling off the end of the function" is the same as if we had written `return None`)
  - once a function returns, that's it—any work done by the function and stored in local variables is lost
  - the next call to the function starts everything anew

- there are times when we'd like to have a "function" which yields a series of values, i.e., it would have to save its state so that the next time it's invoked, it picks up where it left off
  - we use the term "yield" here because in fact we are *not returning* to the caller i.e., we are not returning control of execution to the point where the function was called
  - instead of `return`-ing, we are `yield`-ing, which implies that the transfer of control is temporary and voluntary—our function expects to regain control in the future
- functions that use `yield` instead of `return` are generator functions (or *coroutines* in other languages)
- think of `yield` as `return` + "some magic" for generator functions

## So what's the magic?

- when `yield` is called the state of the generator function is recorded
  - the value of all variables are saved
  - the next line of code to be executed is also saved
  - i.e., the function simply resumes where it left off

In [ ]:

```
def simple_generator():
    yield 1
    yield 'boo!'
    yield 3

for value in simple_generator():
    print(value)
```

## Why do we need generator functions?

- initially they gave programmers an easy way to write code that produced a series of values
  - without generator functions, writing something like a random number generator required a class or module that both generated values and kept track of state between calls
  - with generator functions, doing the above is greatly simplified
- suppose we want a function which, given a list of numbers, returns a list of those numbers which are prime
  - straightforward...

```
def get_primes(nums):
    return ([num for num in nums if is_prime(num)])
```

- now suppose we want to use the above function for very large lists of numbers...so large, in fact, that they won't fit in memory
  - so now we want the function to take a starting value, and return all the primes that are greater than that value
  - since functions only return once, they only have one "chance" to return a value (or list of values)
  - what if our function could return the *next* value, rather than a list?
    - we wouldn't need to create a list at all!

## What is a generator function?

- defined like a normal function, but whenever it needs to generate a value, it does so with the **yield** keyword rather than **return**
  - if the body of a def contains **yield**, the function automatically becomes a generator function (even if it also has a **return** statement)
  - ...there's nothing else we need to do to create one
- generator functions create *generator iterators* (or simply, a *generator*)
  - a generator is a special type of iterator (meaning it has a **\_\_next\_\_()** function)
  - to get the next value from a generator, we use the same built-in function as for iterators: **next()**
- let's return to the more basic notion of a generator function...

### Now we can rewrite our `get_primes()` function as a generator...

```
def get_primes(num):
    while True:
        if is_prime(num):
            yield num
        num += 1
```

- note that if a generator function calls **return** (or simply hits the end of the function), then a **StopIteration** exception is raised, signaling the generator is exhausted (just as an iterator does)

In [ ]:

```
def fibonacci():
    ...
defined as a normal function, but...
...no return keyword

The yield keyword returns a value, but the function retains its state
```

```

...
    prev, curr = 0, 1
    while True:
        yield curr
        prev, curr = curr, prev + curr

f = fibonacci()
print(next(f), next(f), 'before the for loop', sep='\n')

import random

for num in range(0, random.randint(10, 100)):
    val = next(f)
    print(val, end=' ')

```

```

1
1
before the for loop
2 3 5 8 13 21 34 55 89 144 233 377 610 987 1597 2584 4181 6765 10946 17711 28657 46368 75025 121393 196418 317811 514229
832040 1346269 2178309 3524578 5702887 9227465 14930352 24157817 39088169

```

## PEP-342: sending values into generators

- PEP-342 added support for passing values *into* generators using the `send()` function
- let's go back to the prime number example but instead of simply printing every prime number greater than some number, we'll find the smallest prime number greater than successive powers of a number (i.e. for 10, we want the smallest prime greater than 10, then 100, then 1000, etc.)

```

def get_primes(num):
    while True:
        if is_prime(num):
            num = yield num
        num += 1

```

- the `yield` line now says "**yield num, and when a value is sent to me, set num to that value**"
- and we can print the next prime greater than 10, 100, 1000, as follows:

```

def print_successive_primes(iterations, base=10):
    prime_generator = get_primes(base)
    prime_generator.send(None)
    for power in range(iterations):
        print(prime_generator.send(base ** power))

```

- printing `generator.send()` is possible because `send` both sends a value to the generator and returns the value yielded by the generator
- note that the first time we send a value into a generator, it must be `None`

## Now let's look at a generator *expression*

- generator equivalent of a list comprehension

```
In [ ]: squares = [num * num for num in range(1, 11)] # List comprehension  
squares
```

```
In [ ]: squares = {num * num for num in range(1, 11)} # set comprehension  
squares
```

```
In [ ]: squares = {num: num * num for num in range(1, 11)} # dict comprehension  
squares
```

```
In [ ]: squares = (num * num for num in range(1, 11)) # generator expression (NOT a 'tuple comprehension')  
squares
```

```
In [ ]: next(squares), next(squares)
```

```
In [ ]: list(squares) # for thing in squares: print(thing)
```

## Lab: Generators

- modify your random iterator to be a generator function

## itertools

- module of functions for efficient looping

- all of its functions return iterators
- some produce finite sequences
- others produce infinite sequences

In [ ]:

```
from itertools import zip_longest
list1 = ['a', 'b', 'c', 'd']
list2 = ['apple', 'banana', 'cherry']

for item1, item2 in zip_longest(list1, list2, fillvalue='***'):
    print(item1, '=>', item2)
```

```
a => apple
b => banana
c => cherry
d => ***
```

In [ ]:

```
from itertools import count
counter = count(start=789)
for _ in range(10):
    print(next(counter))
```

```
789
790
791
792
793
794
795
796
797
798
```

In [ ]:

```
from itertools import count

counter = count(1, 0.25)
for _ in range(10):
    print(next(counter))
```

```
1
1.25
1.5
1.75
```

```
2.0  
2.25  
2.5  
2.75  
3.0  
3.25
```

```
In [ ]:  
from itertools import cycle  
sizes = ['S', 'M', 'L']  
sc = cycle(sizes)  
#next(sc), next(sc), next(sc), next(sc), next(sc)  
for num in range(1, 22):  
    print(next(sc), end=' ')
```

```
S M L S M L S M L S M L S M L S M L S M L
```

```
In [ ]:  
from itertools import islice  
list(islice(fibonacci(), 50, 60))
```

```
Out[ ]: [20365011074,  
32951280099,  
53316291173,  
86267571272,  
139583862445,  
225851433717,  
365435296162,  
591286729879,  
956722026041,  
1548008755920]
```

```
In [ ]:  
from itertools import count, islice  
  
for num in islice(count(1, 0.25), 13, 17):  
    print(num)
```

```
4.25  
4.5  
4.75  
5.0
```

```
In [ ]:  
# some produce a finite sequence from an infinite sequence  
from itertools import islice, cycle  
colors = cycle(['red', 'white', 'blue']) # infinite
```

```
limited = islice(colors, 0, 5)

for color in limited:
    print(color, end= ' ')
```

```
red white blue red white
```

```
In [ ]: from itertools import chain
rank = list(range(2, 11))
#picture = { 'J': 'Jack', 'Q': 'Queen', 'K': 'King', 'A': 'Ace' }
picture = list('JQKA')

#for card in chain(rank, picture):
#    #print(card, end=' ')

list(chain(rank, picture))
```

```
Out[ ]: [2, 3, 4, 5, 6, 7, 8, 9, 10, 'J', 'Q', 'K', 'A']
```

```
In [ ]: from itertools import filterfalse
# filter out items for which predicate is False
numbers = [7, 12, 20, 23, 32, 44]
list(filterfalse(lambda x: x % 2, numbers))
```

```
Out[ ]: [12, 20, 32, 44]
```

```
In [ ]: # filters elements, returning only those that have a corresponding
# element that evaluates to True
from itertools import compress
words = ['how', 'now', 'brown', 'cow']
counts = [13, '', 'x', None]
list(compress(words, counts))
```

```
Out[ ]: ['how', 'brown']
```

```
In [ ]: # accumulate sums, or other binary functions
from itertools import accumulate
list(accumulate([3, 5, 10, 21]))
#help(accumulate)
```

```
Out[ ]: [3, 8, 18, 39]
```

```
In [ ]: list(accumulate(range(1, 10), lambda x, y: x * y))
```

```
Out[ ]: [1, 2, 6, 24, 120, 720, 5040, 40320, 362880]
```

```
In [ ]: # Amortize a 5% loan of 1000 with 4 annual payments of 250  
cashflows = [1000, 250, 250, 250, 250]  
list(accumulate(cashflows, lambda bal, pmt: bal * 1.05 - pmt))
```

```
Out[ ]: [1000, 800.0, 590.0, 369.5, 137.97500000000002]
```