# Iterating over data with Python



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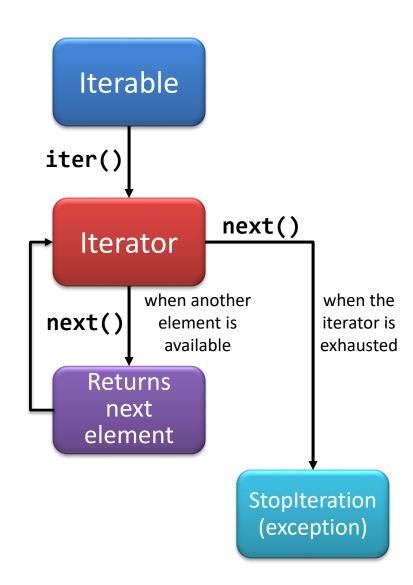
#### Outline

- Iterables and iterators
- Generators
- Comprehensions
- Enumerate and Zip
- Map, Filter and Reduce
- $\lambda$ -functions

#### **ITERABLES AND ITERATORS**

#### **Iterables**

- Iterables are containers of data in which we can iterate to obtain elements one by one
- Lists, tuples, sets, dictionaries, strings, etc. are iterables
- Iterators are objects used to iterate over iterables
- Two important functions:
  - iter(): creates an iterator from an iterable
  - next(): returns the next item



## Iterables: example

```
>>> lst = [1, 2, 3] # lst is an iterable
>>> it = iter(lst) # it is an iterator
>>> next(it)
>>> next(it)
                           next() raises a StopIteration
                           exception when no more items
>>> next(it)
                           are available
3
>>> next(it)
Traceback (most recent call last):
  File "<stdin>", line 1, in <module>
StopIteration
>>>
```

## next(it, default)

```
it = iter(some iterable) # Creates an iterable
# next(it, default) does not raise any exception.
# Instead, it returns the default value.
v = next(it, None)
while v is not None:
    do something(v)
    v = next(it, None)
# Equivalent code
```

for v in it:

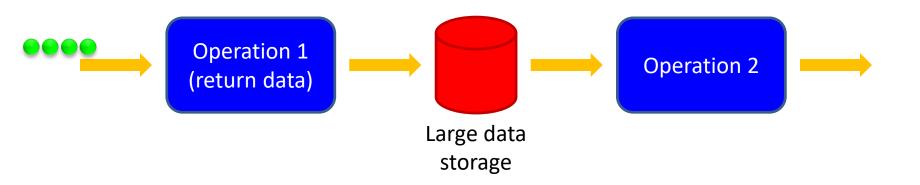
do something(v)

## **GENERATORS**

## Designing data pipelines

 Big data systems often have to process long streams of data with pipelines chaining different operations.

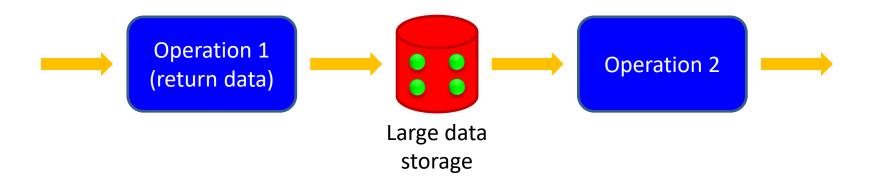
How to store the data between operations?



## Designing data pipelines

 Big data systems often have to process long streams of data with pipelines chaining different operations.

How to store the data between operations?



#### Generators

 A mechanism to avoid storage of large amounts of data.

 Generators are lazy iterators that do not store the whole data structures in memory.



## Generators: example

```
>>> def natural_numbers():
        n = 0
   while True:
            yield n
            n += 1
>>> gen = natural_numbers()
>>> next(gen)
0
>>> next(gen)
>>> next(gen)
2
>>> next(gen)
```

```
>>> for i in natural_numbers():
        if is_prime(i):
            print(i)
3
5
672131
672137
672143
^CTraceback (...)
KeyboardInterrupt
>>>
```

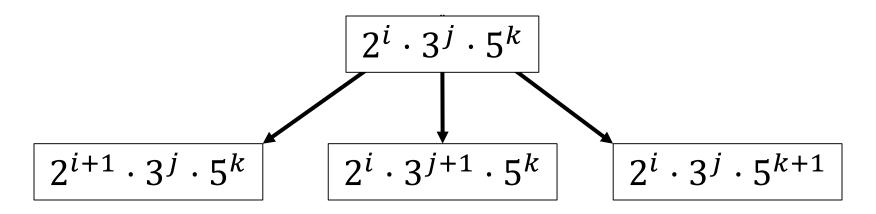
## Hamming numbers

 Hamming numbers are those numbers whose only prime divisors are 2, 3, and 5.

- Examples:
  - 20 is a Hamming number  $(2^2 \cdot 3^0 \cdot 5^1)$
  - 21 is not a Hamming number  $(3 \cdot 7)$
- Exercise: design a program that prints the n smallest Hamming numbers.
- Strategy: for every Hamming number  $2^i \cdot 3^j \cdot 5^k$ , we can generate three new numbers by increasing each one of the exponents.
- How to generate them in ascending order?

## Hamming numbers

Generating the next Hamming numbers:



- How to generate them in ascending order?
  - Use a priority queue to store the pending numbers

- How to avoid repetitions?
  - Remember the last delivered number

# Hamming numbers: simulation

Value	Priority Queue
	1
1	<u>2 3 5</u>
2	3 <u>4</u> 5 <u>6</u> <u>10</u>
3	4 5 6 <u>6</u> <u>9</u> 10 <u>15</u>
4	5 6 6 <u>8</u> 9 10 <u>12</u> 15 <u>20</u>
5	6 6 8 9 10 <b>10</b> 12 15 <b>15</b> 20 <b>25</b>
6	6 8 9 10 10 12 <u>12</u> 15 <u>18</u> 20 25 <u>30</u>
6	8 9 10 10 12 12 15 18 20 25 30
8	9 10 10 12 12 15 <u>16</u> 18 20 <u>24</u> 25 30 <u>40</u>
9	10 10 12 12 15 16 18 <u>18</u> 20 24 25 <u>27</u> 30 40 <u>45</u>
10	10 12 12 15 16 18 18 20 <u>20</u> 24 25 27 30 <u>30</u> 40 45 <u>50</u>
:	•

## Hamming numbers

```
import heapq
from typing import Iterator
def hamming_numbers() -> Iterator[int]:
    """Generates all Hamming numbers in ascending order"""
    pq = [1] # priority queue storing Hamming numbers
    prev = 0 # the last delivered number
    while True:
        value = heapq.heappop(pq) # get the smallest number
        if value != prev: # avoid repetitions
            yield value # deliver the value and wait (lazy)
            prev = value
            for x in 2, 3, 5: # generate new numbers
                nxt = x*value
                heapq.heappush(pq, nxt)
```

## Hamming numbers

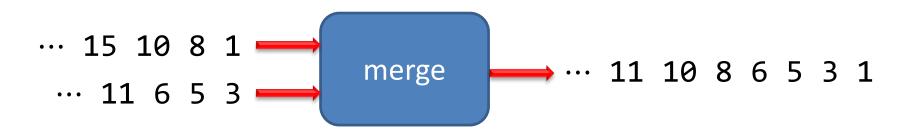
```
def main(n: int) -> None:
    """Test to print first n hamming numbers"""
    hammings = hamming_numbers() # the generator
    for _ in range(n):
        print(next(hammings))

if __name__ == '__main__':
    main(20)
```

## Merging sequences

 Functions can receive iterators as parameters and generate iterators as results

 Let us design a function that merges two sorted iterators and generated a sorted iterator



## Merging sequences: typing

```
from typing import Iterator, TypeVar, Protocol
# This is an abstract class that contains the __lt__
# operator (<). No need to implement it.</pre>
class Comparable(Protocol):
    def __lt__(self: 'T', other: 'T') -> bool: ...
# This is a generic type. The bound attribute indicates
# that the type must contain the operators of Comparable.
T = TypeVar('T', bound=Comparable)
# The merge function dealing with sequences of elements
# that are "Comparable" (i.e., the type contains the
# the operator <).</pre>
def merge(a: Iterator[T], b: Iterator[T]) -> Iterator[T]:
    """reads two sorted iterators and generates a sorted
       iterator by merging them"""
```

## Merging sequences: code

```
def merge(a: Iterator[T], b: Iterator[T]) -> Iterator[T]:
     "reads two sorted iterators and generates a sorted
       iterator by merging them"""
    x, y = next(a, None), next(b, None)
    while x is not None and y is not None:
        if x < y:
            yield x
            x = next(a, None)
        else:
            yield y
            y = next(b, None)
    if x is not None:
        yield x
        yield from a # delivers values from another iterator
    if y is not None:
        yield y
        yield from b
```

#### **COMPREHENSIONS**

## Comprehensions

Set builder notation. Example:

$$S = \{x^2 \mid x < 1000, x \text{ is prime}\}\$$

Conventional Python using a for loop:

```
s = {}
for x in range(1000):
    if is_prime(x):
        s.add(x**2)
```

Using comprehensions:

```
s = \{x**2 \text{ for } x \text{ in } range(1000) \text{ if } is_prime(x)\}
```

## Nested comprehensions

```
# Given a list of words, create a dictionary with the key-value
# pairs <word: number of vowels>
words = ['cat', 'kangaroo', 'lion', 'dog', 'hippopotamus']
# We can use s.count(x) to count the number of occurrences
# of x in the string s, e.g., 'kangaroo'.count('o') is 2
vowels = {w: sum(w.count(x) for x in 'aeiou') for w in words}
print(vowels)
{'cat': 1, 'kangaroo': 4, 'lion': 2, 'dog': 1, 'hippopotamus': 5}
# Let us print a list of the words with more than 3 vowels
print([w for w in vowels.keys() if vowels[w] > 3])
['kangaroo', 'hippopotamus']
```

## Creating matrices with comprehensions

```
# Let us create a 4x4 identity matrix
matrix = [[0]*4]*4
for i in range(4):
    matrix[i][i] = 1
# Surprise! What's wrong?
print(matrix)
[[1, 1, 1, 1], [1, 1, 1, 1], [1, 1, 1, 1], [1, 1, 1, 1]]
# Let us use comprehensions
matrix = [[1 if i==j else 0 for j in range(4)] for i in range(4)]
print(matrix)
[[1, 0, 0, 0], [0, 1, 0, 0], [0, 0, 1, 0], [0, 0, 0, 1]]
# How to create a zero matrix with n rows and m columns
matrix = [[0]*m for _ in range(n)]
```

## Generator expressions: example

```
>>> import sys
>>> # A list comprehension
>>> squares_lc = [i**2 for i in range(10**6)]
>>> # It generates a long list (larger than 8Mb)
>>> sys.getsizeof(squares_lc)
8448728
>>>
>>> # But we can also create a generator using (...)
>>> squares_gc = (i**2 for i in range(10**6))
>>> sys.getsizeof(squares gc)
104
>>> # and we can iterate over the generator
>>> for n in squares_gc:
        if is_prime(n+1):
           print(n+1)
```

#### **ENUMERATE AND ZIP**

#### enumerate

```
# Different ways of printing indices and values
lst = [x**2 for x in range(100)]
for i in range(len(lst)):
    print(i, lst[i])
i = 0
for v in 1st:
    print(i, v)
    i += 1
for i, v in enumerate(lst):
    print(i, v)
# It also works for generators!
lst\_gen = (x**2 for x in range(100))
for i, v in enumerate(lst_gen):
    print(i, v)
```

## Hamming numbers

```
def main(n: int) -> None:
    """Test to print first n hamming numbers"""
    for i, x in enumerate(hamming_numbers()):
        if i == n:
            break
        print(x)
  __name__ == '__main__':
    main(20)
```

## zip

```
>>> # Zipping lists
>>> lst1 = [x**2 for x in range(100)]
>>> lst2 = [2*x for x in range(100)]
>>> lst3 = zip(lst1, lst2)
                                              lst1
                                                       1st2
>>> # 1st3 is an iterator!
>>> print(lst3)
<zip object at 0x7fd8d8beacc0>
>>> for x, y in 1st3:
        print(x, y)
16 8
                                                   1st3
25 10
36 12
49 14
```

## zipping and unzipping

```
>>> # Let us zip two lists
>>> letters = ['a', 'b', 'c', 'd']
>>> numbers = [1, 2, 3, 4]
>>> ln_zip = zip(letters, numbers)
>>> list_ln = list(ln_zip)
>>> print(list_ln)
[('a', 1), ('b', 2), ('c', 3), ('d', 4)]
# Now we can unzip the list of tuples
>>> lett, numb = zip(*list ln)
>>> print('letters =', lett)
letters = ('a', 'b', 'c', 'd')
>>> print('numbers =', numb)
numbers = (1, 2, 3, 4)
```

# MAP, FILTER AND REDUCE

## map, filter and reduce

- Loops, comprehensions and generators are techniques used to process data in iterable objects.
- The functions map(), filter() and reduce() provide a functional programming approach to achieve similar goals.
- They can be applied to any iterable object (list, tuple, set, ...)
- These functions can provide a very elegant solution to compute expressions like this:

$$\sum_{0 \le i < n, \text{ is\_prime}(i)} i^2$$

## map, filter, reduce: auxiliary functions

```
def square(x: int) -> int:
    return x*x
def add(x: int, y: int) -> int:
    return x + y
def is_prime(n: int) -> bool:
    if n <= 1:
        return False
    d = 2
    while d*d <= n:
        if n%d == 0:
            return False
        d += 1
    return True
```

```
lst = [1, 2, 3, 4, 5, 6]
# map creates an iterator that applies a function
# to all elements of the iterable object
result = map(square, 1st)
list(result)
# Output: [1, 4, 9, 16, 25, 36]
```

#### filter

```
# filter creates an iterator that selects the
# elements that satisfy the filtering condition
result = filter(is_prime, range(30))

list(result)
# Output: [2, 3, 5, 7, 11, 13, 17, 19, 23, 29]
```

#### reduce

#### from functools import reduce

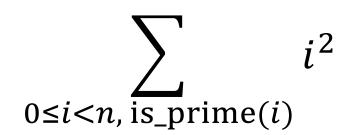
```
# reduce visits all elements and executes a
# function that "accumulates" their values
result = reduce(add, range(10))
result
# Output: 45
# An initial value can also be specified
result = reduce(add, range(10), 5)
result
# Output: 50
```

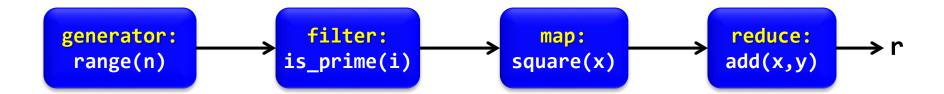
## any, all

**any()** and **all()** are particular cases of reduce functions with Boolean results.

```
>>> numbers = [2, 3, 7, 11, 13, 23]
>>> all(is prime(x) for x in numbers)
True
\Rightarrow \Rightarrow all(x%2 == 1 for x in numbers)
False
\Rightarrow any (6 < x < 12 for x in numbers)
True
>>> words = ['cat', 'kangaroo', 'lion', 'dog', 'hippopotamus']
>>> all(len(w) > 10 for w in words)
False
>>> any(len(w) > 10 for w in words)
True
>>> all(len(w) < 15 for w in words)</pre>
True
>>> any(w[0] == 'h' for w in words)
True
```

## Back to our problem



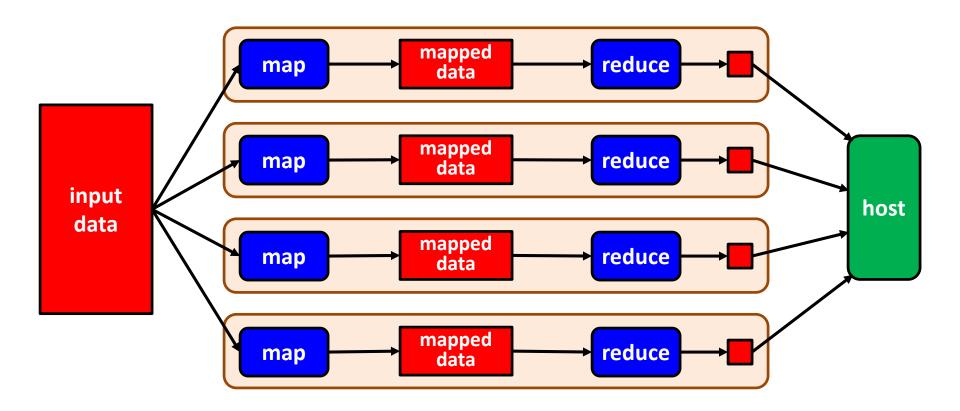


```
n = 10**7
r = reduce(add, map(square, filter(is_prime, range(n))))
print(r)
# Output: 21113978675102768574
```

**Important:** no intermediate lists are generated. Very low storage is required (< 1000 bytes). By using lists to store the intermediate results, about 400Mb of storage would be required.

# MapReduce

A programming model for big data sets using parallel, distributed algorithms



# **λ-FUNCTIONS**

#### $\lambda$ -functions

 Lambda functions are anonymous functions that receive parameters and return expressions

• Syntax:

```
lambda parameters: expression
```

is equivalent to:

```
def anonymous(parameters):
    return expression
```

• Examples:

```
lambda x: x*x
lambda x, y: x+y
```

# Using $\lambda$ -functions in map/filter/reduce

```
\sum_{0 \le i < n, \text{ is_prime}(i)} i^2
```

#### Conclusions

 Processing long streams of data is one of the main tasks of big data systems. Memory storage is one of the critical resources

 When designing data pipelines exploit lazy evaluation mechanisms to generate data upon demand and avoid unnecessary data storage

### **EXERCISES**

# Comprehensions

 Create a dictionary where the keys are the numbers of a list and the values are the highest one-digit divisor of each number

• Create a list with all positive numbers smaller than n that are divisible by some number included in a list called divisors

 Given a rectangular matrix (list of lists), calculate its transpose using list comprehensions

# Generating the Fibonacci series

- Design a generator of the Fibonacci series
- Given a list of divisors, design a generator that generates the Fibonacci numbers that are divisible by all divisors of the list. Example:

```
divisors = [3, 5, 7, 11]
gen = (...) # design the generator
for x in gen:
    print(x)

Output:
0
102334155
23416728348467685
5358359254990966640871840
1226132595394188293000174702095995
```

## Intersection of sequences

Implement the function intersect with the following specification:

```
from typing import Iterator, TypeVar, Protocol
class Comparable(Protocol):
    def __lt__(self: 'T', other: 'T') -> bool: ...
T = TypeVar('T', bound=Comparable)
def intersect(a: Iterator[T], b: Iterator[T]) -> Iterator[T]:
    """reads two sorted iterators and generates a sorted
       iterator with only the common elements"""
# Example:
      a = [1, 3, 5, 5, 6, 7, 10, 13, 16, 18]
      b = [2, 5, 5, 8, 13, 13, 15, 16, 20]
      output: [5, 13, 16]
```

### Farey sequence

• The Farey sequence of order n is the sequence of completely reduced fractions between 0 and 1 with denominators less than or equal to n, arranged in ascending order. Example:

$$F_5 = \left\{ \frac{0}{1}, \frac{1}{5}, \frac{1}{4}, \frac{1}{3}, \frac{2}{5}, \frac{1}{2}, \frac{3}{5}, \frac{2}{3}, \frac{3}{4}, \frac{4}{5}, \frac{1}{1} \right\}$$

 Design the generator farey(n) that generates the Farey sequence of order n:

```
def farey(n: int) -> Iterator[tuple[int, int]]:
```

- Write Python expressions to calculate:
  - the sum of the elements of  $F_n$
  - the number of elements of  $F_n$

Hint: The next element of the Farey sequence can be calculated using only the two previous elements (find the rule in Wikipedia!)

# Filter/reduce pipeline

Design two versions of the following function using a filter-reduce pipeline:

- One version with auxiliary functions
- One version with lambda functions

```
from dataclasses import dataclass
from typing import Iterable
from functools import reduce
@dataclass
class Person:
    name: str
    age: int
    salary: float
def avg salary(people: Iterable[Person],
               min_age: int, max_age: int) -> float:
    """Returns the average salary of the people with age
       between min_age and max_age"""
```