# A Look at the Python Collections Module

Roger Hurwitz March 30, 2023 Boston Python User Group

#### Special thanks to:

- Emily CharlesNed Batchelder



#### Who am I?

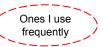
- Senior Python software director/developer at Intel
- Web-based server system apps/services focus
- Pre-Python, I programmed in C/C++ for years
- A few of my favorite things about Python:
  - The language: intuitive and elegant (to me)
  - o The ecosystem: "batteries included"
  - The web: best in class (backend) framework support
  - The people: "I came for the language and stayed for the community" (Brett Cannon)



#### **Collections Overview**

- Collections module is ~18 years old, first showing up in Python 2.4
- The module "implements specialized container datatypes providing alternatives to Python's general purpose built-in containers, dict, list, set, and tuple"
- Hot Take: While many datatypes in the module are still quite useful, others are showing their age
- Further Reading:
  - https://docs.python.org/3/library/collections.html
  - http://pvmotw.com/2/collections/
  - o https://realpython.com/python-collections-module/

namedtuple()	factory function for creating tuple subclasses with named field			
deque	list-like container with fast appends and pops on either end			
ChainMap	dict-like class for creating a single view of multiple mappings			
Counter	dict subclass for counting hashable objects			
OrderedDict	dict subclass that remembers the order entries were added			
defaultdict	dict subclass that calls a factory function to supply missing values			
UserDict	wrapper around dictionary objects for easier dict subclassing			
UserList	wrapper around list objects for easier list subclassing			
UserString	wrapper around string objects for easier string subclassing			



# Collections in a Nutshell

Datatype	Utility	Key Feature	Alternatives	Hot Take	
(namedtuple)	Med	Adds named fields to tuple but still interchangeable w/ tuple	dataclass, tuple, typing.NamedTuple	Great readability substitute for tuple, but dataclass better for other cases	
deque	Med	Much higher performance than list for element insert/remove intensive use cases	list	deque features make it well suited for queue, stack, and buffer operations	
ChainMap	Med/Low	Single virtual dict composed of multiple actual dict objects	dict.update	Situational, but very useful when data structured with nested scopes/contexts	
Counter	Med	Has multiple methods in support of counting operations on iterables	pandas, dict, defaultdict,	Counter is viable for bare bones data analysis but lacks scalability/flexibility	
OrderedDict	Low	As of Python 3.7, less about key ordering and more about order-based methods	dict	Situational, but equality operator and reordering performance can matter	
(defaultdict)	Med/High	Caller specified default factory function automatically invoked when adding keys	dict.setdefault	Subtle: unlike setdefault, its object factory only invoked on key creation	
User: Dict, List, String	Med/Low	Specifically designed to be better than base types for subclassing	dict, str, list	With Python duck typing (protocols), less needed than in other languages	

# Examples

## namedtuple

#### **Enhance readability**

```
from collections import namedtuple

pivMod = namedtuple("DivMod", ["quotient", "remainder"])

def readable_divmod(x, y):
    """Return built-in divmod result as namedtuple."""
    dm = divmod(x, y)
    return DivMod(quotient=dm[0], remainder=dm[1])

result = readable_divmod(10, 3)
    print(f"{result = }")
    print(f"{result.quotient = }")
    print(f"{result.remainder = }")

result = DivMod(quotient=3, remainder=1)
result.quotient = 3
result.remainder = 1

... without losing the benefits of tuple
```

```
result = readable_divmod(10, 3)
print(f"{isinstance(result, tuple) = }")
print(f"{result[0] = }, {result[1] = }")
quotient, remainder = result
print(f"{quotient = }, {remainder = }")

isinstance(result, tuple) = True
result[0] = 3, result[1] = 1
quotient = 3, remainder = 1
```

# deque

```
The name of the game is "speed"

1 from collections import deque
```

#### ...but there's more

# Be careful who you trust...



How is that better than using append on a fixed length deque? Doesn't appending to a fixed length deque push the oldest element out?



You're right, appending to a fixed length 'deque' object will indeed push out the oldest element. However, using the 'rotate' method has a couple of advantages over simply appending to a fixed length deque:



1

- 1. Efficiency: When you append an element to a `deque` object, it has to shift all the elements over by one position to make room for the new element. This can be an expensive operation, especially if the `deque` is large. In contrast, the `rotate` method simply rotates the elements of the `deque` in place, which can be a more efficient operation, especially if the number of positions to rotate is small.
- 2. Flexibility: Using the `rotate` method allows you to implement more complex circular buffer behaviors than simply appending to a fixed length `deque`. For example, you might want to rotate the buffer by a different number of positions depending on some condition, or you might want to rotate the buffer in both directions. The `rotate` method gives you this flexibility.

That being said, if you only need to implement a simple circular buffer and efficiency is not a concern, using 'deque.append' to overwrite the oldest element will work just fine.

2



I don't agree with your efficiency assertion. deque, unlike list, uses links to connect elements therefore it should not require any element shifting



You are correct. I apologize for the mistake in my previous response. 'deque' does indeed use a doubly-linked list to connect its elements, so appending an element to a fixed-length 'deque' should not require any element shifting.





# ChainMap

#### Scope-Based Lookup

```
1 from collections import ChainMap
 3 member specials = {
        "Chicken": 6.79,
        "Wine": 3.64,
 6 }
 7 monthly_specials = {
        "Butter": 1.49.
 9
        "Wine": 4.64,
10 }
11 regular_prices = {
      "Butter": 1.99.
     "Chicken": 8.79.
     "Wine": 5.64.
14
       "Bread": 2.99,
15
16 }
17
18 member prices = ChainMap(member specials, monthly specials, regular prices)
19
20 print(f"{member prices['Wine'] = }")
21 print(f"{member prices['Butter'] = }")
22 print(f"{member_prices['Bread'] = }")
member_prices['Wine'] = 3.64
member prices['Butter'] = 1.49
member_prices['Bread'] = 2.99
```

#### **Shortcut for Skipping First Mapping**

```
1 nonmember_prices = member_prices.parents
2
3 print(f"{nonmember_prices['Wine'] = }")
4 print(f"{nonmember_prices['Butter'] = }")
5 print(f"{nonmember_prices['Bread'] = }")
nonmember_prices['Wine'] = 4.64
nonmember_prices['Butter'] = 1.49
nonmember_prices['Bread'] = 2.99
```

#### Counter

#### Base Case: Count the Elements in an Iterable

```
from collections import Counter
import random

# create a list of random coin flip results
coinflip_outcomes = ["heads", "tails"]
coinflips = random.choices(coinflip_outcomes, k=100)

coinflip_cnt = Counter(coinflips)
print(f"{coinflip_cnt = }")

coinflip_cnt = Counter({'tails': 59, 'heads': 41})
```

#### Next Level: Slice and Dice the Results

```
print(f"{coinflip_cnt.most_common(1) = }")
print(f"{coinflip_cnt.total() = }")
print(f"{[elem for elem in coinflip_cnt.elements()] = }")
print(f"{coinflip_cnt['edges'] = }")

coinflip_cnt.most_common(1) = [('tails', 59)]
coinflip_cnt.total() = 100
[elem for elem in coinflip_cnt.elements()] = ['tails', 'tails', '
```

#### OrderedDict

#### In Python 3.6 dict Found Order

```
from collections import OrderedDict

ordered_dict = OrderedDict.fromkeys("abcdef")

regular_dict = dict.fromkeys("abcdef")

for keys in zip(ordered_dict, regular_dict):
    print(keys, end=" ")

('a', 'a') ('b', 'b') ('c', 'c') ('d', 'd') ('e', 'e') ('f', 'f')
```

#### Equality: OrderedDict vs dict

```
ordered_dict_a = OrderedDict.fromkeys("abcdef")
ordered_dict_b = OrderedDict.fromkeys("bacdef")

print(f"{ordered_dict_a = ordered_dict_b = }")

ordered_dict_a == ordered_dict_b = False

regular_dict_a = dict.fromkeys("abcdef")
regular_dict_b = dict.fromkeys("bacdef")

print(f"{regular_dict_a = regular_dict_b = }")

regular_dict_a == regular_dict_b = True
```

#### Reordering Flexibility/Efficiency

```
ordered_dict = OrderedDict.fromkeys("abcdef")

ordered_dict.move_to_end("f", last=False)

for key in ordered_dict:
    print(key, end=" ")

f a b c d e
```

#### defaultdict

```
from collections import defaultdict

pets = [
    ("cat", "Luna"), ("cat", "Lily"), ("bird", "Bella"), ("cat", "Lucy"),
    ("cat", "Nala"), ("gerbil", "Callie"), ("cat", "Kitty"), ("cat", "Cleo"),
    ("bird", "Willow"), ("cat", "Chloe"), ("dog", "Max"), ("gerbil", "Charlie"),
    ("dog", "Cooper"), ("bird", "Milo"), ("dog", "Buddy"), ("bird", "Rocky"),
    ("dog", "Bear"), ("dog", "Teddy"), ("dog", "Duke"), ("gerbil", "Leo"),
]
```

#### Typical Use: Initialize New Keys w/ List

```
pet_names_by_type = defaultdict(list)

for pet_type, pet_name in pets:
    pet_names_by_type[pet_type].append(pet_name)

print(f"{pet_names_by_type['dog'] = }")

pet_names_by_type['dog'] = ['Max', 'Cooper', 'Buddy', 'Bear', 'Teddy', 'Duke']
```

#### dict Alternative: Less Readable/Efficient

```
pet_names_by_type = {}

for pet_type, pet_name in pets:
    pet_names_by_type.setdefault(pet_type, []).append(pet_name)

print(f"{pet_names_by_type['dog'] = }")

pet_names_by_type['dog'] = ['Max', 'Cooper', 'Buddy', 'Bear', 'Teddy', 'Duke']
```

# Exercises

## Customers.csv

#### https://github.com/rogerhurwitz/pythonCollectionsPresentation

	customer_id	gender	age	annual_income	spending_score	profession	work_experience	family_size
1	1	Male	19	15000	39	Healthcare	1	
2	2	Male	21	35000	81	Engineer	3	
3	3	Female	20	86000	6	Engineer	1	
4	4	Female	23	59000	77	Lawyer	0	
5	5	Female	31	38000	40	Entertainment	2	
6	6	Female	22	58000	76	Artist	0	
7	7	Female	35	31000	6	Healthcare	1	
8	8	Female	23	84000	94	Healthcare	1	
9	9	Male	64	97000	3	Engineer	0	
10	10	Female	30	98000	72	Artist	1	
11	11	Male	67	7000	14	Engineer	1	
12	12	Female	35	93000	99	Healthcare	4	
13	13	Female	58	80000	15	Executive	0	
14	14	Female	24	91000	77	Lawyer	1	
15	15	Male	37	19000	13	Doctor	0	
16	16	Male	22	51000	79	Healthcare	1	
17	17	Female	35	29000	35	Homemaker	9	
18	18	Male	20	89000	66	Healthcare	1	
19	19	Male	52	20000	29	Entertainment	1	
20	20	Female	35	62000	98	Artist	0	
21	21	Male	35	96000	35	Homemaker	12	
22	22	Male	25	4000	73	Healthcare	3	
23	23	Female	46	42000	5	Artist	13	
24	24	Male	31	71000	73	Artist	5	
25	25	Female	54	67000	14	Executive	1	
26	26	Male	29	52000	82	Artist	1	
27	27	Female	45	68000	32	Healthcare	9	

# Exercise 0: Execute this on your computer

```
import csv
    with open("Customers.csv", newline="") as csvfile:
        customer reader = csv.reader(csvfile)
        headers = next(customer reader)
        customers = [
            tuple(row)
            for row in customer reader
 11 print(f"{customers[0] = }")
customers[0] = ('1', 'Male', '19', '15000', '39', 'Healthcare', '1', '4')
```

# Exercise 1: Replace use of tuple with namedtuple ¶

```
from collections import namedtuple
   import csv
   with open("Customers.csv", newline="") as csvfile:
       customer reader = csv.reader(csvfile)
       headers = next(customer_reader)
       Customer = namedtuple("Customer", headers)
       customers = [
           Customer(*row)
           for row in customer_reader
14
   print(f"{customers[0] = }")
```

customers[0] = Customer(customer id='1', gender='Male', age='19', annual income='15000', spending

# Exercise 2: Use Counter and show the three most common customer professions

```
from collections import namedtuple, Counter
import csv

with open("Customers.csv", newline="") as csvfile:
    customer_reader = csv.reader(csvfile)
    headers = next(customer_reader)

Customer = namedtuple("Customer", headers)

profession_cnt = Counter([
    Customer(*row).profession
    for row in customer_reader
])

profession_cnt.most_common(3)
```

[('Artist', 612), ('Healthcare', 339), ('Entertainment', 234)]

### Exercise 3: Clean this code using collections module

```
1 from collections import namedtuple
   import csv
   with open("Customers.csv", newline="") as csvfile:
       customer reader = csv.reader(csvfile)
       headers = next(customer reader)
 6
       Customer = namedtuple("Customer", headers)
 9
10
       genders by profession = {}
11
12
       for row in customer reader:
           customer = Customer(*row)
13
14
15
           if genders by profession.get(customer.profession) is None:
               genders by profession[customer.profession] = {}
16
17
18
           if genders by profession[customer.profession].get(customer.gender) is None:
               genders by profession[customer.profession][customer.gender] = 1
19
20
           else:
21
               genders by profession[customer.profession][customer.gender] += 1
22
23 print(genders by profession)
```

```
{'Healthcare': {'Male': 143, 'Female': 196}, 'Engineer': {'Male': 76, 'Female': 103}, 'Lawyer': {'Female': 86, 'Ma 2}, 'Executive': {'Female': 87, 'Male': 66}, 'Doctor': {'Male': 72, 'Female': 89}, 'Homemaker': {'Female': 39, 'Ma
```

```
import csv
    with open("Customers.csv", newline="") as csvfile:
        customer reader = csv.reader(csvfile)
  5
        headers = next(customer reader)
  8
        Customer = namedtuple("Customer", headers)
 9
        genders by profession = defaultdict(Counter)
10
11
12
        for row in customer reader:
13
            customer = Customer(*row)
            genders by profession[customer.profession][customer.gender] += 1
14
15
16 print(genders by profession)
defaultdict(<class 'collections.Counter'>, {'Healthcare': Counter({'Female': 196, 'Male': 143}), 'Engine
t': Counter({'Female': 133, 'Male': 101}), 'Artist': Counter({'Female': 380, 'Male': 232}), 'Executive':
({'Female': 39, 'Male': 21}), 'Marketing': Counter({'Female': 53, 'Male': 32}), '': Counter({'Female': 2
```

from collections import namedtuple, defaultdict, Counter