# Looping using the .iterrows() function

OPTIMIZING PYTHON CODE WITH PANDAS



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# The poker dataset

	<b>S1</b>	R1	<b>S2</b>	R2	<b>S</b> 3	R3	<b>S4</b>	R4	<b>S5</b>	R5
1	1	10	3	11	3	13	4	4	2	1
2	2	11	2	13	2	10	2	12	2	1
3	3	12	3	11	3	13	3	10	3	1

- 1. Hearts
- 2. Diamonds
- 3. Clubs
- 4. Spades

# **Generators in Python**

```
def city_name_generator():
    yield('New York')
    yield('London')
    yield('Tokyo')
    yield('Sao Paolo')

city_names = city_name_generator()
```

```
next(city_names)
'New York'
next(city_names)
'London'
next(city_names)
'Tokyo'
next(city_names)
'Sao Paolo'
next(city_names)
Traceback (most recent call last):
  File "<stdin>", line 1, in <module>
```

StopIteration

# Looping using the .iterrows() function

```
gen = poker.iterrows()
first_element = next(gen)
first_element[0]
0
first_element[1]
      13
      12
Name: 1, dtype: int64
```

# Using the .iterrows() function

```
start_time = time.time()
for index, values in range(poker.shape[0]):
    next
print("Time using range(): {} sec".format(time.time() - start_time))
```

Results using range(): 0.006870031 sec

```
data_generator = poker.iterrows()

start_time = time.time()

for index, values in data_generator:
    next

print("Time using .iterrows(): {} sec".format(time.time() - start_time))
```

Time using .iterrows(): 1.55003094673 sec

# Let's do it!

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# Looping using the .apply() function

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# The .apply() function

```
data_sqrt = poker.apply(lambda x: np.sqrt)
head(data_sqrt, 4)
```

```
I S1
            | R1
                        1 S2
                                     R2
                                                           R3
                                               S3
1.000000 | 3.162278 | 1.000000 | 3.316625
                                           | 3.464102 |
                                                       1.000000
1.414214 |
           3.316625 | 1.414214 | 3.605551
                                            1.414214
                                                        3.162278
           3.464102 | 1.732051 | 3.316625
1.732051
                                            1.732051
                                                        3.605551
           3.162278 | 2.000000 | 3.316625 | 2.000000
2.000000 |
                                                        1.000000
```

```
data_sqrt_2 = np.sqrt(poker)
```



# The .apply() function for rows

```
apply_start_time = time.time()
poker[['R1', 'R2', 'R3', 'R4', 'R5']].apply(lambda x: sum(x), axis=1)
print("Time using .apply(): {} sec".format(time.time() - apply_start_time))
```

```
Time using .apply(): 0.636334896088 sec
```

```
start_time = time.time()
for ind, value in poker.iterrows():
    sum([value[1], value[3], value[5], value[7], value[9]])
print("Time using .iterrows(): {} sec".format(time.time() - start_time))
```

```
Time using .iterrows(): 3.15526986122 sec
```

```
Difference in speed: 395.85051529%
```

# The .apply() function for columns

```
start_time = time.time()
poker[['R1', 'R2', 'R3', 'R4', 'R5']].apply(lambda x: sum(x), axis=0)
print("Time using .apply(): {} sec".format(time.time() - apply_start_time))
```

```
Time using .apply(): 0.00490880012 seconds
```

```
start_time = time.time()
poker[['R1', 'R1', 'R3', 'R4', 'R5']].sum(axis=0)
print("Time using pandas: {} sec".format(time.time() - start_time))
```

```
Time using pandas: 0.00279092788 sec
```

```
Difference in speed: 160.310951649%
```



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# Vectorization over Pandas series

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# DataFrames as arrays

#### not vectorized

vectorized

a b
1 \* 6

2 \* 7

3 \* 8

4 \* 9

5 \* 10

5 operations

b

a

1 6

2 | | 7

3 8 4 9

5 \* 10

2 operations

### How to perform pandas vectorization

```
start_time = time.time()
poker[['R1', 'R2', 'R3', 'R4', 'R5']].sum(axis=1)
print("Time using pandas vectorization: {} sec".format(time.time() - start_time))
Time using pandas vectorization: 0.0026819705 sec
poker[['R1', 'R2', 'R3', 'R4', 'R5']].sum(axis=1).head()
 dtype: int64 | -- |
```

## Comparison to the previous methods

```
data_generator = data.iterrows()

start_time = time.time()

for index, value in data_generator:
        sum([value[1], value[3], value[7]])

print("Time using .iterrows(){} seconds" % (time.time() - start_time))
```

Results from the above operation calculated in 3.37918996 seconds

```
start_time = time.time()
data[['R1', 'R2', 'R3', 'R4', 'R5']].apply(lambda x: sum(x),axis=1)
print("Results from the above operation calculated in %s seconds" % (time.time() - start_time))
```

Results from the above operation calculated in 0.637711048 seconds

```
- Difference between vectorization and the `.iterows()` function: 111,800.75%
- Difference between vectorization and the `.apply()` function: 20,853%
```



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# Vectorization with NumPy arrays using .values()

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# NumPy in pandas

```
df = pd.DataFrame({'Col1':[0, 1,
2, 3, 4, 5, 6]}, dtype=np.int8)
print(df)
```

```
nd = np.array(range(7))
print(nd)

[0 1 2 3 4 5 6]
```

## How to perform NumPy vectorization

```
start_time = time.time()
poker[['R1', 'R2', 'R3', 'R4', 'R5']].values.sum(axis=1)
print("Time using NumPy vectorization: {} sec(time.time() - start_time))
```

Results from the above operation calculated in 0.00157618522644 seconds

```
start_time = time.time()
poker[['R1', 'R2', 'R3', 'R4', 'R5']].sum(axis=1)
print("Results from the above operation calculated in %s seconds" % (time.time() - start_time))
```

Results from the above operation calculated in 0.00268197059631 seconds

```
Difference in time: 39.0482%
```



# Let's do it!

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