# A brief, non-conclusive overview of some good coding practices

#### Follow Style Guidelines

- PEP 8: <a href="https://peps.python.org/pep-0008/">https://peps.python.org/pep-0008/</a>
- Google Style Guide: <a href="https://google.github.io/styleguide/pyguide.html">https://google.github.io/styleguide/pyguide.html</a>
- If you code using an IDE, you can install linters and code formatters to help you with this. (<u>flake8</u>, <u>black</u>, ...)

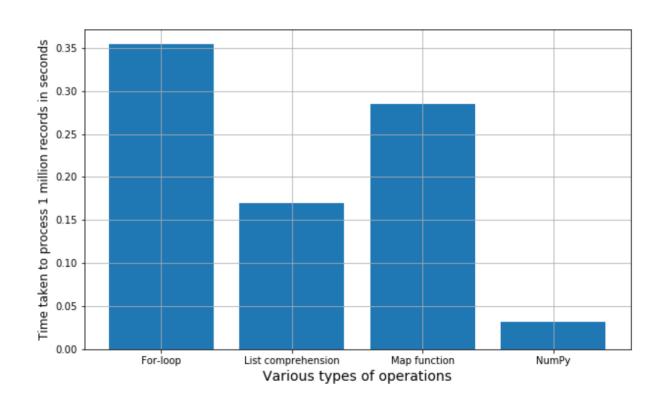
## Docstrings, type hinting, exceptions and comments

```
def cov(X, rowvar = False):
    X = X.clone()
    if X.dim() < 2:
        X = X.view(1, -1)
    if not rowvar and X.size(0) != 1:
        X = X.t()
    fact = 1.0 / (X.size(1) - 1)
    X -= torch.mean(X, dim=1, keepdim=True)
    Xt = X.t()
    return fact * X.matmul(Xt).squeeze()</pre>
```

```
def cov(X: Tensor. rowvar: bool = False) -> Tensor:
    """Estimate a covariance matrix given data.
    Covariance indicates the level to which two variables vary together.
    If we examine N-dimensional samples, X = [x 1, x 2, ... x N]^T,
    then the covariance matrix element `C {ij}` is the covariance of
    `x i` and `x j`. The element `C {ii}` is the variance of `x i`.
    Args:
        X: A 1-D or 2-D array containing multiple variables and observations.
            Each row of `X` represents a variable, and each column a single
            observation of all those variables.
        rowvar: If `rowvar` is True, then each row represents a
            variable, with observations in the columns. Otherwise, the
            relationship is transposed: each column represents a variable,
            while the rows contain observations.
    Returns:
        The covariance matrix of the variables.
   X = X.clone()
    if X.dim() > 2:
        raise ValueError("m has more than 2 dimensions")
    if X.dim() < 2:
       X = X.view(1. -1)
    if not rowvar and X.size(0) != 1:
       X = X.t()
    # m = m.tvpe(torch.double) # uncomment this line if desired
    fact = 1.0 / (X.size(1) - 1)
    X -= torch.mean(X, dim=1, keepdim=True)
   Xt = X.t() # if complex: mt = m.t().conj()
    return fact * X.matmul(Xt).squeeze()
```

#### Vectorization

```
[1e1, 1e2, 1e3]
12 =
t1 = time.time()
for item in 11:
    12.append(lg10(item))
t2 = time.time()
speed.append(t2-t1)
VS.
   = np.array(11)
   = time.time()
v2 = np.log10(v1)
t2 = time.time()
speed.append(t2-t1)
```



From: https://medium.com/productive-data-science/why-you-should-forget-for-loop-for-data-science-code-and-embrace-vectorization-696632622d5f

#### Vectorization

VS.

```
C = A@B
```

```
C = []
for i in range(len(A)):
    row = []
    for j in range(len(B[0])):
        product = 0
        for v in range(len(A[i])):
            product += A[i][v] * B[v][j]
        row.append(product)
    C.append(row)
```

#### Loops

Loop over elements rather than index of elements for items in

```
list
                 1 = [0, 1, 2, 3]
                 for i in range(len(l)):
                      print(l[i])
                 VS.
                 for item in 1:
                      print(item)
                 for i, item in enumerate(1):
                     print(i, item)
```

#### Loops

Use zip to iterate over multiple lists

```
11 = [0, 1, 2, 3]
12 = [3, 2, 1, 0]
for i in range(len(l1)):
    print(l1[i], l2[i])
VS.
for i, j in zip(11, 12):
    print(i, j)
```

#### Loops

#### Looping over dictionaries

```
d = \{"a": 1, "b": 2, "c": 3\}
for key in d.keys():
    print(key)
VS.
for key in d:
    print(key)
for key, val in d.items():
    print(key, val)
```

#### Conditionals

```
If cond == True:
    pass

vs.

if cond:
    pass
```

#### Strings

```
result = 1.23456789123456789
print("The result is " + str(result) + " units.")
VS.
print("The result is %s units." %result)
print(f"The result is {result} units.")
print("The result is {:.2f} units.".format(result)) # prints only
2 decimals
```

#### List/Dict comprehensions

```
11 = ["0","1","2","3"]
12 = [0,1,2,3]
d = {}
for key, val in zip(l1,l2):
    d[key] = val

vs.

d = {key: val for key, val in zip(l1, l2)}
```

#### Context managers

```
f = open("text.txt", "r"):
text = f.read()
f.close()

vs.

with open("text.txt", "r") as f:
   text = f.read()
```

### Type checking

```
if type(x) == float:
    pass

vs.

if isinstance(x, float):
    pass
```

```
if x == None:
    pass

vs.

if x is None:
    pass
```

#### Notation

```
x = 100000
y = 0.00001

VS.

x = 1e5
y = 1e-5
```

## The Walrus (for Python >= v3.8)

```
while (user_answer := input(f"\n{question} ")) not in valid_answers:
    print(f"Please answer one of {', '.join(valid_answers)}")

[value for num in numbers if (value := slow(num)) > 0]
```

#### **Avoid**

From module import \*

-Overwrites functions that are already within the namespace, i.e. from numpy import \* will overwrite sum() with numpy.sum().

# Version control and collaborative coding

Consider working with Git / GitHub

## Avoid merge conflicts with jupyter notebooks

- `nbdev` provides git hooks that clean notebooks of unecessary metadata and prevent your notebooks from braking during merge conflicts.
- you can check out how to install it here:

github.com/jnsbck/nbdev\_demo.git