# Data Science With Python

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# Contents

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# Overview

These are some notes that may be helpful for computational biology analysis that focuses on Python use.

# Python

#### 2.1 General

Here is a general skeleton that can be used to start a python script that takes input.

```
#!/usr/bin/env python

def runArgparse():
    import argparse

parser = argparse.ArgumentParser()
    parser.add_argument('--indir', '-i', type=str, nargs='*', help='Input directory containing the vcf file
    parser.add_argument('--loadolddata', '-o', action='store_true', help='Load previously existing data.')

args = parser.parse_args()
    indir = args.indir
    return indir

if __name__ == '__main__':
    runArgparse()
```

#### 2.2 Numba

Numba speeds up python code without having to switch to a different interpreter, and doesn't require static typing of variables as Cython does. Just calling Numba will increase the speed of a script (except during the compilation which will add some time). But this isn't the best way to take advantage of the speed boost.

Here is an example script that uses jit to invoke Numba.

```
#!/usr/bin/env python

from numba import jit
import numpy as np
import time

def go_slow(x):
    for i in range(14):
```

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```
x *= x
@jit(nopython=True)
def go_fast(x):
   for i in range(14):
       x *= x
# DO NOT REPORT THIS... COMPILATION TIME IS INCLUDED IN THE EXECUTION TIME!
start = time.time()
go_slow(5)
end = time.time()
print("Elapsed slow (with compilation) = %s" % (end - start))
start = time.time()
go_fast(5)
end = time.time()
print("Elapsed fast (with compilation) = %s" % (end - start))
# DO NOT REPORT THIS... COMPILATION TIME IS INCLUDED IN THE EXECUTION TIME!
start = time.time()
go_slow(5)
end = time.time()
print("Elapsed slow (after compilation) = %s" % (end - start))
start = time.time()
go_fast(5)
end = time.time()
print("Elapsed fast (after compilation) = %s" % (end - start))
```

Parallelization can be used to automatically utilize multiple cores.

```
from numba import njit, prange
@njit(parallel=True)
def prange_test(A):
    s = 0
    for i in prange(A.shape[0]):
        s += A[i]
    return s
```

### 2.3 Cython

Installation

```
conda install -c conda-forge cython
```

Ipython usage

```
%load_ext Cython

%%cython
def f(x):
    return 2 * x
or

def f(int x):
    return 2 * x
```

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timeit(f(4))

# JupyterLab

Here is a simple template that I use that controls a couple useful things when starting a new notebook.

```
import sys
sys.path.append('../util')

%reload_ext autoreload
%autoreload 2

from util import *
import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
import seaborn as sns

sns.set_palette('pastel')
sns.set_style('ticks')
sns.set_context('paper', font_scale=1)
```

It is often convenient to have a notebook automatically refresh the imported libraries so that they can be modified while working on a JupyterLab notebook.

```
%reload_ext autoreload
%autoreload 2
```

To allow directory organization, depended on the separated into different directories and imported into a jupyter notebook using the following import statement.

```
import sys
sys.path.append('../util')
```

A table of contents can be created to refer to each of the headers throughout a notebook in html format. The code is below (Obviously needs to be simplified.)

```
<span><a href="#Change-the-width-of-the-page" data-toc-modified-id="Change-the-width-of-the-pag"</pre>
       <span><a href="#Import-packages" data-toc-modified-id="Import-packages-1.2"><span class="to</pre>
<1i>
   <span><a href="#Colours" data-toc-modified-id="Colours-2"><span class="toc-item-num">2&nbsp;&nb
<span><a href="#Colour-line-graph" data-toc-modified-id="Colour-line-graph-2.1"><span class="to</pre>
   <span><a href="#Totals-for-studies" data-toc-modified-id="Totals-for-studies-3"><span class="to</pre>
<span><a href="#Functions-for-calculating-trinucleotide-context-specific-mutation-rates" data-t</pre>
   <span><a href="#Calculating-mutation-rates-for-individual-variants" data-toc-modified-id="C</pre>
       <span><a href="#DNMT3A" data-toc-modified-id="DNMT3A-4.1.1"><span class="toc-item-num">4.1.
       <span><a href="#TET2" data-toc-modified-id="TET2-4.1.2"><span class="toc-item-num">4.1.
       <span><a href="#ASXL1" data-toc-modified-id="ASXL1-4.1.3"><span class="toc-item-num">4.
       <span><a href="#TP53" data-toc-modified-id="TP53-4.1.4"><span class="toc-item-num">4.1.
   <span><a href="#Calculating-mutation-rates-from-a-.csv-file-of-variants" data-toc-modified-</pre>
       <span><a href="#DNMT3A" data-toc-modified-id="DNMT3A-4.2.1"><span class="toc-item-num">
       <span><a href="#TET2" data-toc-modified-id="TET2-4.2.2"><span class="toc-item-num">4.2.
       <span><a href="#ASXL1" data-toc-modified-id="ASXL1-4.2.3"><span class="toc-item-num">4.
       <span><a href="#TP53" data-toc-modified-id="TP53-4.2.4"><span class="toc-item-num">4.2.
   <span><a href="#Calculating-mutation-rates-from-a-list-of-variants" data-toc-modified-id="C</pre>
       <span><a href="#DNMT3A" data-toc-modified-id="DNMT3A-4.3.1"><span class="toc-item-num">
   <span><a href="#TET2" data-toc-modified-id="TET2-4.3.2"><span class="toc-item-num">4.3.2&nb
       <span><a href="#ASXL1" data-toc-modified-id="ASXL1-4.3.3"><span class="toc-item-num">4.
       <span><a href="#TP53" data-toc-modified-id="TP53-4.3.4"><span class="toc-item-num">4.3.
<span><a href="#Lists-of-variants-targeted-by-each-study" data-toc-modified-id="Lists-of-varian"</pre>
   <span><a href="#Genovese-2014" data-toc-modified-id="Genovese-2014-5.2"><span class="toc-it</pre>
   <span><a href="#McKerrel-2015" data-toc-modified-id="McKerrel-2015-5.3"><span class="toc-it"</pre>
   <span><a href="#Zink-2017" data-toc-modified-id="Zink-2017-5.4"><span class="toc-item-num">
   <span><a href="#Coombs-2017" data-toc-modified-id="Coombs-2017-5.5"><span class="toc-item-n"</pre>
   <span><a href="#Young-2016-&amp;-2019" data-toc-modified-id="Young-2016-&amp;-2019-5.6"><sp</pre>
   <span><a href="#Desai-2018" data-toc-modified-id="Desai-2018-5.7"><span class="toc-item-num"</pre>
   <span><a href="#Acuna-Hidalgo-2017" data-toc-modified-id="Acuna-Hidalgo-2017-5.8"><span cla</pre>
<span><a href="#Lists-of-all-possible-variants-in-DNMT3A,-TET2,-ASXL1,-TP53" data-toc-modified-</pre>
   <span><a href="#DNMT3A" data-toc-modified-id="DNMT3A-6.1"><span class="toc-item-num">6.1&nb
   <span><a href="#TET2" data-toc-modified-id="TET2-6.2"><span class="toc-item-num">6.2&nbsp;&
   <span><a href="#ASXL1" data-toc-modified-id="ASXL1-6.3"><span class="toc-item-num">6.3&nbsp
<span><a href="#TP53" data-toc-modified-id="TP53-6.4"><span class="toc-item-num">6.4&nbsp;&nbsp
<span><a href="#Actual-number-of-observations-of-each-variant" data-toc-modified-id="Actual-num"</pre>
   <1i>>
       <span><a href="#DNMT3A" data-toc-modified-id="DNMT3A-7.0.1"><span class="toc-item-num">
       <span><a href="#TET2" data-toc-modified-id="TET2-7.0.2"><span class="toc-item-num">7.0."
   <span><a href="#ASXL1" data-toc-modified-id="ASXL1-7.0.3"><span class="toc-item-num">7.0.3&
```

# Visualization

#### 4.1 Color

#### 4.1.1 Colorschemes

```
Seaborn Themes

Pastel: {'Blue':'#a3c6ff', 'Orange':'#f7ab60', 'Green':'#60f7a9', 'Red':'#fc9d94', 'Purple':'#bea3ff',

Deep: {'Green':'#5baf68'}
```

#### 4.1.2 Controlling Coloration

Not all plots automatically plot with a white background, and when using something dark like jupyterlab or a presentation this can be frustrating. The background color can be set in pyplot like this.

```
fig.patch.set_facecolor('xkcd:mint green')
```

When plotting, samples will not always be colored with the same color, especially when different subsets of samples are included in different plots. Here is a manual workaround to specify the coloration of displayed data. This is a bit cumbersome so there might be a more elegant way of achieving the same outcome.

#### 4.2 Matplotlib

Plotting a heatmap.

```
import matplotlib.pyplot as plt
import numpy as np
a = np.random.random((16, 16))
plt.imshow(a, cmap='RdBu'', interpolation='nearest')
plt.show()
```

Possible heatmap colors are:

Accent, Accent\_r, Blues, Blues\_r, BrBG, BrBG\_r, BuGn, BuGn\_r, BuPu, BuPu\_r, CMRmap, CMRmap\_r, Dark2, Da Set1\_r, Set2, Set2\_r, Set3, Set3\_r, Spectral, Spectral\_r, Wistia, Wistia\_r, YlGn, YlGnBu, YlGnBu\_r, YlGgist\_stern, gist\_stern\_r, gist\_yarg, gist\_yarg\_r, gnuplot, gnuplot2, gnuplot2\_r, gnuplot\_r, gray, gray\_twilight, twilight\_r, twilight\_shifted, twilight\_shifted\_r, viridis, viridis\_r, vlag, vlag\_r, winter, w

A simple venn diagram.

```
from matplotlib_venn import venn2
venn2(subsets = (3, 2, 1))
```

A more complicated venn diagram.

```
from matplotlib import pyplot as plt
import numpy as np
from matplotlib_venn import venn3, venn3_circles
plt.figure(figsize=(4,4))
v = venn3(subsets=(1, 1, 1, 1, 1, 1, 1), set_labels = ('A', 'B', 'C'))
v.get_patch_by_id('100').set_alpha(1.0)
v.get_patch_by_id('100').set_color('white')
v.get_label_by_id('100').set_text('Unknown')
v.get_label_by_id('A').set_text('Set "A"')
c = venn3_circles(subsets=(1, 1, 1, 1, 1, 1, 1), linestyle='dotted')
c[0].set_lw(1.0)
c[0].set_ls('dotted')
plt.title("Sample Venn diagram")
plt.annotate('Unknown set', xy=v.get_label_by_id('100').get_position() - np.array([0, 0.05]), xytext=(-
```

4.2. MATPLOTLIB

An upset plot is a nice alternative to a traditional venn diagram. The project is hosted here, and this is the documentation.

First install the library.

```
pip install upsetplot
```

Here is the code to create the plot.

```
import numpy as np
arrays = [[False,False,False,True,True,True,True],
          [False, False, True, True, False, False, True, True],
          [False, True, False, True, False, True, False, True]]
tuples = list(zip(*arrays))
def o(one=False, two=False, three=False):
        temp = pd.merge(indels[(indels.Individual==one)], indels[(indels.Individual==two)], how='inner'
        return len(pd.merge(temp, indels[(indels.Individual==three)], how='inner', on=['Loc', 'Var']))
        return len(pd.merge(indels[(indels.Individual==one)], indels[(indels.Individual==two)], how='in
    elif one:
        return len(indels[(indels.Individual==one)])
    else:
        return 0
index = pd.MultiIndex.from_tuples(tuples, names=['Ind 1', 'Ind 2', 'Ind 3'])
s = pd.Series([o(),
               o(3),
               o(2),
               o(2,3),
               o(1),
               o(1,3),
               o(1,2),
               o(1,2,3)], index=index)
from upsetplot import plot as up
plt.savefig("../images/indels.svg", format="svg", bbox_inches="tight")
```

Log scales seem to always be a challenge. Here is at least one solution to change ticks to log manually.

```
ax1.set_yticks(y_major_ticks)
ax1.set_yticklabels(y_major_tick_labels, fontsize = axisfont)
ax1.yaxis.set_tick_params(width=scale, color = grey3, length = 6)
```

#### 4.3 Seaborn

Here is a general bar plot that includes some commonly used parameters.

```
# fits my 22 inch monitor
plt.figure(figsize=(19.17,11.98))
# order controls the display order of the samples
sns.catplot(x="Sample", y="Somatic", kind="bar", data=var_counts, order=labels);
# keeps x-axis labels, but eliminates the tick mark
plt.tick_params(labelbottom=True, bottom=False)
# trim off the x-axis
sns.despine(offset=10, trim=True, bottom=True)
# labels
plt.title('')
plt.ylabel('', fontsize=8)
plt.xlabel('', fontsize=8)
# manual control of xlabels
labels = ['Indiv_1-a','Indiv_2','Indiv_3','Indiv_1-b']
# control xtick order
plt.xticks(range(len(labels)), labels, rotation=45)
# control the number of x-ticks
plt.locator_params(axis='x', nbins=10)
# legend positioning
plt.legend(loc='upper right')
# log scale
plt.gca().set_yscale('log')
# this is better if neg values are needed
plt.gca().set_yscale('symlog')
# fit plot to display
plt.tight_layout()
plt.show()
# save figure with tight layout
plt.savefig("test.svg", format="svg", bbox_inches="tight", dpi=1000)
```

Signifiance information can be added by including p-values and label bars using the following code.

```
x1, x2 = 0, 1 # columns to annotate on the plot y2, y1 = 20, 15 # placement of the line and how for down the vertical legs go plt.plot([x1,x1, x2, x2], [y1, y2, y2, y1], linewidth=1, color='k') # stats line plt.text((x1+x2)*.5, y2+2, "p=0.09", ha='center', va='bottom', fontsize=8) # p-value or sig
```

#### 4.4 Statistics

This is a two-sided T-test for the null hypothesis that two populations have the same means. It is important to note that it assumes the population variances are the same, so this must be changed if the assumption is incorrect.

```
# ttest_ind(a, b, axis=0, equal_var=True, nan_policy='propagate')
from scipy.stats import ttest_ind
ttest_ind(df[df['sample'] == 'one']['means'], df[df['sample'] == 'two']['means'])
```

### 4.5 Various Plot Styles

This displays each individual datapoint overlayed on a boxplot

```
ax = sns.boxplot(x='day', y='total_bill', data=tips)
ax = sns.swarmplot(x='day', y='total_bill', data=tips, color='.25')
```

# **Biology**

#### 5.1 General

Some helpful commands for genetic sequence.

```
from string import ascii_uppercase # python 3
from string import upper, lower # python 2
upper('tcga')
lower('TCGA')
title('tcga') # capitalize the first letter
```

### 5.2 Biopython

Reverse complement of sequence

```
from Bio.Seq import Seq
str(Seq(i).reverse_complement())
```

### 5.3 UCSC Genome Browser

Get sequence from UCSC genome browser

```
from subprocess import check_output, STDOUT

temp = check_output('wget -q0- http://genome.ucsc.edu/cgi-bin/das/hg19/dna?segment=%s:%s,%s' % (vcfObj.
```

If the reference genome comes in the .2bit format, it is likely that it should then be converted to .fa format, and .2bittofa can accomplish this.

```
twoBitToFa refgenomes/hg19.2bit refgenomes/hg19.fa
```

The newly created fa file will need to be indexed if it is to be used with gatk and bwa mem.

Here is an example of indexing where -a bwtsw specifies that we want to use the indexing algorithm that is capable of handling the whole human genome.

```
bwa index -a bwtsw reference.fa
```

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Then create a file called reference.fa.fai, with one record per line for each of the contigs in the FASTA reference file. Each record is composed of the contig name, size, location, basesPerLine and bytesPerLine.

```
samtools faidx reference.fa
```

#### 5.4 Ref Genome

Get sequence from reference genome

```
from subprocess import check_output, STDOUT
temp = check_output('samtools faidx %s %s:%s-%s' % (ref, vcf0bj.chrom, low, high), stderr=STDOUT, shell
finalSeq = ''
for line in temp.decode('UTF-8').split('\n'):
for line in temp.decode('UTF-8').split('\n'): # this is only necessary in python 3 to convert binary to
    if '>' not in line:
        finalSeq += line

finalSeq = finalSeq.upper()
```

#### 5.5 Personal Information

```
# parse vcf file with parseline
if '#' not in line and 'chr' in line: # skip the info
# vcf handling
from parseline import VCFObj
# or
from util import VCFObj
vcfObj = VCFObj(vcfLine)
# available attributes: ao, dp, af, wt, var, chrom, location
```

# Data I/O

## 6.1 Reading Data Files

```
Opening .gz files
```

```
import gzip
for line in gzip.open('myFile.gz'):
    print line
```

### 6.2 Pickles

Writing data in pickle format

```
import pickle
p = open('principle.pkl', 'wb')
pickle.dump(principleData, p)
p.close()
```

Reading data in pickle format

```
import pickle
p = open('principle.pkl', 'rb')
principleData = pickle.load(p)
p.close()
```

# **Pandas**

## 7.1 File I/O

Read a csv file into a DataFrame.

```
pd.read_csv(filepath)
Write a DataFrame to a file.
x.to_csv(path_or_buf='outputDir', sep='\n', header=False, index=False)
```

#### 7.2 Data Structure Creation

Create a DataFrame.

```
frame = pd.DataFrame(np.random.randn(4,3), columns=list('bde'), index=['Utah','Ohio','Texas','Oregon'])

A DataFrame can conveniently be created from a dictionary.

import pandas as pd
```

```
import pandas as pd
data = {'AAA' : [4,5,6,7], 'BBB' : [10,20,30,40],'CCC' : [100,50,-30,-50]}
df2 = pd.DataFrame(data=data,index=[1,2,3,4]) #Note index starts at 1.
df2
```

```
## AAA BBB CCC
## 1 4 10 100
## 2 5 20 50
## 3 6 30 -30
## 4 7 40 -50
```

#### 7.3 Selection

Is data within a DataFrame found within a dictionary or list? (Instead of a dictionary a series can be used and maybe another DataFrame)

```
import pandas as pd
df = pd.DataFrame({'A': [1, 2, 3], 'B': ['a', 'b', 'f']})
df.isin([1, 3, 12, 'a'])
```

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```
##
           Α
                   В
## 0
               True
       True
     False False
## 1
## 2
       True False
df[df.isin([1, 3, 12, 'a'])]
##
              В
         Α
      1.0
## 0
              a
## 1
      {\tt NaN}
          {\tt NaN}
## 2 3.0
            NaN
```

Data within a DataFrame can be selected based on position within the DataFrame.

```
import pandas as pd
df2.iloc[1:3]
```

```
## AAA BBB CCC
## 2 5 20 50
## 3 6 30 -30
```

Data within a DataFrame can be selected based on position within the DataFrame.

```
import pandas as pd
df2.loc[1:3]
```

```
## AAA BBB CCC
## 1 4 10 100
## 2 5 20 50
## 3 6 30 -30
```

The opposite of matching data can be selected with the inverse operator.

```
df[~((df.AAA <= 6) & (df.index.isin([0,2,4])))]
```

### 7.4 Splitting

Concatenate two DataFrames together without dropping any values or renaming indices.

```
left = pd.concat([left,left])
```

Concatenate two DataFrames together without dropping values, but renaming index.

```
left = pd.concat([left,left], ignore_index=True)
```

Count the number of each unique value in a specified column.

```
left['key1'].value_counts()
left.key1.value_counts()
```

Value counts can also be calculated as percentages so that raw counts as percent makeup can be compared.

```
left['key1'].value_counts(normalize=True) * 100
```

Two DataFrames can be merged such that only the data containing matching keys is retained.

```
result = pd.merge(left, right, how='inner', on=['key1', 'key2'])
```

This DataFrame merge will retain all of the data in the right DataFrame.

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```
result = pd.merge(left, right, how='right', on=['key1', 'key2'])
```

Filter by multiple columns.

```
df[(df.one == 1) & (df.two == 2)]
```

Filter by multiple columns but only return certain values.

```
# this just returns the data in column AAA

df = pd.DataFrame({'AAA' : [4,5,6,7], 'BBB' : [10,20,30,40],'CCC' : [100,50,-30,-50]})

newseries = df.loc[(df['BBB'] < 25) & (df['CCC'] >= -40), 'AAA']
```

Filtering by values and using assignment will modify the original DataFrame.

```
df.loc[(df['BBB'] > 25) | (df['CCC'] >= 75), 'AAA'] = 0.1
```

Select multiple values from a particular column, where Letter is the column header.

```
df[df.Letter.isin(['a','b'])]
```

Use itertools to find combinations of data within a column of two DataFrames.

```
itertools.product(df1['a'], df2['a'])
```

Add data to a particular cell within a DataFrame.

```
df.loc[index,column]=num
```

Make a copy of a DataFrame.

```
df.copy(deep=True)
```

Iterate through a DataFrame.

```
for i in df.itertuples():
    pass
```

Change order of columns.

```
x = x.reindex(columns=['header','seq','plus','qual'])
```

Make a DataFrame from a dictionary

```
d = {'col1': [1, 2], 'col2': [3, 4]}
x = pd.DataFrame(d)
```

Sample from a DataFrame.

```
df.sample(frac=1)
df.sample(n=20, axis=1)
```

Append to a DataFrame.

```
df=df.append(newdf, ignore_index=True) # without ignore_index, the original indices will be used
```

Remove duplicates

```
x = x[~x.index.duplicated(keep='first')] # most ideal method

data = pd.DataFrame({'k1':['one','two']*3+['two'],'k2':[1,1,2,3,3,4,4]})
data.duplicated() # identify duplicate data
data['k1'].duplicated()
data['k1'].drop_duplicates()
```

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```
data.drop_duplicates['k1'] # this does the same thing as the previous line data.drop_duplicates(['k1','k2'], keep='last') # drops unique found in k1 and k2 and keeps the last ind

Check if string is within strings in a given column
```

### 7.5 Relabeling

Rename a column or group of columns can be done by passing a dictionary of the changes.

```
df = df.rename(columns={'a':'b','c':'d'})
```

### 7.6 Sorting and Arranging

x[x['strLoc'].str.contains(region)]

The data in a DataFrame can be sorted in numeric or lexicographic order. The following code sorts the values within the columns a and b.

```
df.sort_values(['a','b'], ascending=False)
Set a column as the new index
x.set_index(['uniques'])
```

### 7.7 Editing Data

Drop columns from a DataFrame.

Changing the datatype of a column of data can be done by just changing column type.

```
df.Age = df.Age.astype(str)

Replace values.
data = pd.Series([1., -999., 2., -999., -1000., 3.])
data.replace(-999, np.nan)
```

#### 7.7.1 Replace values

New data can be set within a DataFrame one subset at a time in a way that will avoid the SettingWith-CopyWarning.

```
import pandas as pd
df = pd.DataFrame({'Trait':['Seed_Shape', 'Seed_Shape', 'Flower_Color', 'Flower_Color'],
                    'Phenotype':['Round','Wrinkled','Purple','White']})
df.loc[df.Trait == 'Seed_Shape', 'Affected_Part'] = 'Seed'
df.loc[df.Trait == 'Flower_Color', 'Affected_Part'] = 'Flower'
print(df)
##
             Trait Phenotype Affected_Part
        Seed_Shape
                       Round
                                       Seed
## 1
        Seed_Shape
                    Wrinkled
                                       Seed
## 2 Flower_Color
                      Purple
                                     Flower
## 3 Flower_Color
                       White
                                     Flower
```

There is a more simple alternative to the above method buit it may result in the SettingWithCopyWarning.

```
df = df.replace('pork', 'bacon')
```

### 7.8 Combining Data Structures

The following merges df and df2 using inner to get the intersection on the Sample column, where indexes are ignored if the merging is performed on a column as in the following example. The other possible merging strategies are: left: use only keys from left frame, similar to a SQL left outer join; preserve key order. right: use only keys from right frame, similar to a SQL right outer join; preserve key order. outer: use union of keys from both frames, similar to a SQL full outer join; sort keys lexicographically. inner: use intersection of keys from both frames, similar to a SQL inner join; preserve the order of the left keys.

```
df = pd.merge(df, df2, how='inner', on=['Sample'])
```

Appending to a Dataframe attaches a DataFrame after another one.

```
df = pd.DataFrame([[1, 2], [3, 4]], columns=list('AB'))
df2 = pd.DataFrame([[5, 6], [7, 8]], columns=list('AB'))
df.append(df2)
```

## 7.9 Summarizing

The mean of column values can be calculated where each of the columns is grouped by the data in a specified column.

```
temp[['Sample','VAF','Var_Count']].groupby('Sample').mean()
```

## 7.10 Arithmetic and Row/Column-wise Analysis

Sometimes it is helpful to analyze the value in a particular cell in a conditional manner depending on it's value and then set the result of this analysis to a corresponding cell in a new column. Here is an example where the VAF of a variant is conditionally analyzed.

```
def LOH(x):
    if x > 0.75: return 1 - x
    elif x <= 0.75 and x > 0.25: return abs(0.5 - x)
    else: return 0
```

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```
all_vars['LOH'] = all_vars.VAF.transform(LOH)
max_loh = all_vars.groupby('Sample').LOH.max().reset_index().rename(columns={'LOH':'Max_LOH'})
all_vars = pd.merge(all_vars, max_loh, how='inner', on=['Sample'])
```

Broadcasting arithmetic is an efficient method of calculating across an entire DataFrame.

```
frame = pd.DataFrame(np.arange(12.).reshape((4,3)), columns=list('bde'), index=['Utah','Ohio','Texas','
series = frame.iloc[0]
frame - series
# the subtraction function could also be used
# frame.sub(series, axis='columns')
```

Apply a function to each row or column.

```
f = lambda x: x.max() - x.min()
frame.apply(f, axis='index')
```

Add two sets of data together, and use fill\_value to avoid replacing any missing data with NaN.

```
x = pd.DataFrame([1,2,3], columns=list('0'))
y = pd.DataFrame([1,2,3], columns=list('1'))
x = x.add(y, fill_value=0)
```

Take the mean or std across specified columns and append as a new column. Below the DataFrame has columns 1-7 that will be used in computing the mean or std and this new data will be appended in a new column labeled 'Mean' or 'Std'.

```
x['Mean']=x[[1,2,3,4,5,6,7]].mean(axis=1)
x['Std']=x[[1,2,3,4,5,6,7]].std(axis=1)
```

# Git

### 8.1 Setup

#### 8.1.1 Git Setup

The username and email needs to be added after git is installed.

```
git config --global user.name "me"
git config --global user.email "me@gmail.com"
```

After this information has been set, it can be checked.

```
git config --list
```

#### 8.1.2 Repository Initiation

To setup a repository, create a folder with an initial file like a README and then initiate it.

```
git init
git status
```

#### 8.1.3 Mirror on Online Repository

Create a repository on a repository like github, gitlab, bitbucket, or sourceforge. Then the local git repository can be synched with the online repository.

```
git remote add origin url-of-online-repository-here
git push -u origin master
```

Of course the repository could just be setup first and then cloned.

```
git clone url-of-online-repository-here
```

CHAPTER 8. GIT

### 8.2 Manipulating Commits

#### 8.2.1 Repository Status

The commit history of a repository can be displayed in verbose form and in summarized form.

```
git log git log --oneline
```

#### 8.2.2 File Checkout

To restore a previous version of a file it can be checked out by first identifying the version to be used using the log history and then restoring the desired file.

```
git log --oneline
git checkout <commit number> file.txt
```

#### 8.2.3 Resetting a Repository

To discard the effect of the previous operation on a file.

```
git reset HEAD file.txt
```

The previous version of the a file can then be restored.

```
git checkout -- file.txt
```

# $\overline{ ext{VIM}}$

### 9.1 Formatting

Automatic newlines are inserted by default; this behavior can be overidden with the following.

```
:set wrap
:set textwidth=0 wrapmargin=0
```

### 9.2 Spellcheck

To setup spellchecking first setup a personal dictionary file.

```
# make a directory for personal dictionary
mkdir -p ~/.vim/spell/
```

Then refer to the dictionary file within VIM, and enable spellchecking.

```
# set personal dictionary
:set spellfile=~/.vim/spell/en.utf-8.add
# turn spellcheck on
:set spell
```

Get spellcheck commands.

```
:help spell
```

Add a word to personal dictionary.

```
--
```

Move to next and previous misspelled word.

```
]s
[s
```

Get suggestions for misspelled word.

```
z=
```

CHAPTER 9. VIM

## Web APIs

#### 10.1 Ensembl

The Ensembl Rest API has a number of different genomics tools.

Here is an example where the Rest API is used to get the genomic locus and amino acid change using only the protein name and amino acid position and identities.

```
content-type=application/json
wget -q --header='Content-type:application/json' 'https://rest.ensembl.org/map/translation/ENSMUSP00000
```

Alternatively python can be used to make the same call.

```
import requests, sys

server = "https://rest.ensembl.org"
ext = "/map/translation/ENSMUSP00000020991/878...879?"

r = requests.get(server+ext, headers={ "Content-Type" : "application/json"})

if not r.ok:
    r.raise_for_status()
    sys.exit()

decoded = r.json()
print(repr(decoded))
```

Here are examples getting variant effect consequences of a particular mutation.

```
wget -q --header='Content-type:application/json' 'https://rest.ensembl.org/vep/mus_musculus/hgvs/ENSMUS
```

#### 10.2 UCSC Genome Browser

Here is an example where human genomic regions are converted to the orthologous mouse regions. First the human sequence is obtained.

```
wget -0 - http://genome.ucsc.edu/cgi-bin/das/hg38/dna?segment=%s:%s,%s >> locs\n' % (chrom, low, high)
```

Then the above sequence is used in UCSC BLAT to find the orthologous region within mouse.

# Golang

#### 11.1 Installation

Installation of linuxbrea

```
sh -c "$(curl -fsSL https://raw.githubusercontent.com/Linuxbrew/install/master/install.sh)"

# Add to path

test -d ~/.linuxbrew && eval $(~/.linuxbrew/bin/brew shellenv)

test -d /home/linuxbrew/.linuxbrew && eval $(/home/linuxbrew/.linuxbrew/bin/brew shellenv)

test -r ~/.bash_profile && echo "eval \$($(brew --prefix)/bin/brew shellenv)" >>~/.bash_profile

echo "eval \$($(brew --prefix)/bin/brew shellenv)" >>~/.profile

# debian/ubuntu dependencies

sudo apt-get install build-essential curl file git
```

LinuxBrew golang installation

```
brew install go
```

The GO PATH should then be checked; it should typically exist at  $\sim$ /go but it can be checked like this echo \$GOPATH

## 11.2 Updating

```
brew install dep
brew upgrade dep
brew cask install spotify
```

## 11.3 Sample Program

Create a file called hello.go

```
package main
import "fmt"
```

```
func main() {
    fmt.Printf("hello, world\n")
}

The program can just be run with
go run main.go

Then compile the program
    go build

Importing multiple things can be done on one line separated by a semicolon
    import ("fmt"; "math")
```

### 11.4 Type conversion

```
Check the type of a variable
import("fmt";"reflect")
fmt.Println(reflect.TypeOf(i))
Convert int to float64
var i int = 5
j := float64(i)
Convert float64 to in
x := 4.0
int(x)
int to string
s := strconv.Itoa(97) // s == "97"
int64 to string
var n int64 = 97
s := strconv.FormatInt(n, 10) // s == "97" (decimal) in base 10
string to int
s := "97"
if n, err := strconv.Atoi(s); err == nil {
    fmt.Println(n+1)
} else {
    fmt.Println(s, "is not an integer.")
}
string to int64
s := "97"
n, err := strconv.ParseInt(s, 10, 64)
if err == nil {
    fmt.Printf("%d of type %T", n, n)
}
```

int to int64