```
'''This cell adapted from:
https://colab.research.google.com/drive/1DofKEdQYaXmDWBzuResXWWvxhLgDeVyl#scrollTo=0hy
To use the Kaggle API, sign up for a Kaggle account at
https://www.kaggle.com. Then go to the 'Account' tab of
your user profile (https://www.kaggle.com/<username>/account)
and select 'Create API Token'. This will trigger the download
of kaggle.json, a file containing your API credentials.
# Run this cell and select the kaggle.json file downloaded
# from the Kaggle account settings page.
from google.colab import files
files.upload()
# Next, install the Kaggle API client.
!pip install -q kaggle
# The Kaggle API client expects this file to be in ~/.kaggle,
# so move it there.
!mkdir -p ~/.kaggle
!cp kaggle.json ~/.kaggle/
# This permissions change avoids a warning on Kaggle tool startup.
!chmod 600 ~/.kaggle/kaggle.json
# List available datasets.
!kaggle datasets list
```

Choose Files | kaggle.json

 kaggle.json(application/json) - 70 bytes, last modified: 6/11/2019 - 100% done Saving kaggle.json to kaggle.json ref title

chicago/chicago-copa-cases himanshupoddar/zomato-bangalore-restaurants sfinspiredu/synchrotron-data-set crisparada/brazilian-cities taniaj/australian-election-2019-tweets romainpessia/artificial-lunar-rocky-landscape-dataset gqfiddler/scotus-opinions se18m502/bee-hive-metrics brittabettendorf/berlin-airbnb-data PromptCloudHQ/world-happiness-report-2019 jlesuffleur/granddebat thegurus/spanish-high-speed-rail-system-ticket-pricing leomauro/smmnet snocco/missing-migrants-project austinreese/craigslist-carstrucks-data robseidl/tennis-atp-tour-australian-open-final-2019 cityofLA/los-angeles-traffic-collision-data inIT-OWL/versatileproductionsystem alvarob96/spanish-stocks-historical-data

Chicago COPA Cases Zomato Bangalore Restau Synchrotron Data Set Brazilian Cities Australian Election 201 Artificial Lunar Landsc SCOTUS Opinions Beehive Metrics Berlin Airbnb Data World Happiness Report Le Grand Débat National Spanish High Speed Rail SMMnet Missing Migrants Projec Craigslist Cars+Trucks Tennis ATP Tour Austral Los Angeles Traffic Col Versatile Production Sy Spanish Stocks Historic mfekadu/darpa-timit-acousticphonetic-continuous-speech DARPA TIMIT Acoustic-Ph

```
Do not run them.'''
from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call d

cd drive/My\ Drive/Datasets

C→ /content/drive/My Drive/Datasets

mkdir Kaggle_HIV

cd Kaggle_HIV/

/content/drive/My Drive/Datasets/Kaggle_HIV

!kaggle competitions download -c hivprogression

ls

Гэ

T> test data.csv training data.csv

import pandas as pd

#Look at the data we've got:

data = pd.read_csv('training_data.csv')
data.head()

	PatientID	Resp	PR Seq	
0	1	0	CCTCAAATCACTCTTTGGCAACGACCCCTCGTCCCAATAAGGATAG	CCC
1	2	0	CCTCAAATCACTCTTTGGCAACGACCCCTCGTCGCAATAAAGATAG	CC(
2	3	0	CCTCAAATCACTCTTTGGCAACGACCCCTCGTCGCAATAAAGGTAG	CC(
3	4	0	CCTCAAATCACTCTTTGGCAACGACCCCTCGTCGCAATAAGGATAG	CC(
4	5	0	CCTCAAATCACTCTTTGGCAACGACCCCTCGTCGCAGTAAAGATAG	CC(

```
#What are the columns?
data.columns
```

Index(['PatientID', 'Resp', 'PR Seq', 'RT Seq', 'VL-t0', 'CD4-t0'], dtype='obje

#How is the data distributed?
data['Resp'].value_counts()

Name: Resp, dtype: int64

#How do the sequences vary? This looks at length:

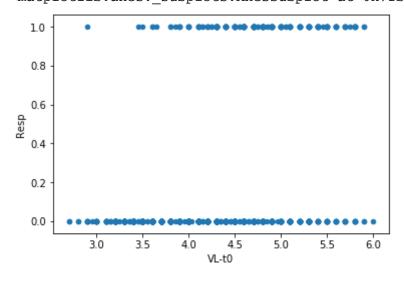
data['PR Seq'].dropna().apply(lambda x: len(str(x))).value_counts()

```
297
        889
294
          14
267
           6
270
           3
285
           2
252
           2
276
           1
261
           1
255
           1
216
           1
Name: PR Seq, dtype: int64
```

#How is Viral Load and Response linked?

data.plot.scatter('VL-t0', 'Resp')

C→ <matplotlib.axes. subplots.AxesSubplot at 0x7fb6478960b8>



#The above plot looks like the relationship follows
#a logistic regression - this is common for binary
#data (given Resp is either 0 or 1, we call the data binary)

#Run a simple Logistic regression to predict

```
#responsiveness from Viral Load and CD4 count:
#(You can ignore the warnings this generates, if any.)

from sklearn.linear_model import LogisticRegression

model = LogisticRegression(solver='lbfgs')
X = data[['VL-t0', 'CD4-t0']]
y = data['Resp']
model.fit(X,y)
y_pred = model.predict(X)
acc = sum(data['Resp']==y_pred)/len(data['Resp'])
print('Accuracy of Logistic Regression Model: {}%'.format(acc*100))
```

Accuracy of Logistic Regression Model: 78.7%