# **Automating API Pull using Python**

# (1) Import Python Libraries

#### In [1]:

```
from requests import Request, Session
from requests.exceptions import ConnectionError, Timeout, TooManyRedirects
import json

import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

import os
from time import time
from time import sleep
```

# (2) Using an API Key with Python

#### In [3]:

```
url = 'https://pro-api.coinmarketcap.com/v1/cryptocurrency/listings/latest'

parameters = {
    'start':'1',
    'limit':'15',
    'convert':'USD'
}

headers = {
    'Accepts': 'application/json',
    'X-CMC_PRO_API_KEY': '0d11cefa-2730-441a-b86e-d5bc32d407d7',
}
```

#### In [4]:

```
session = Session()
session.headers.update(headers)
```

#### In [5]:

```
try:
    response = session.get(url, params=parameters)
    data = json.loads(response.text)
    print(data)
vexcept (ConnectionError, Timeout, TooManyRedirects) as e:
    print(e)
```

{'status': {'timestamp': '2023-08-18T07:38:54.184Z', 'error code': 0, 'error\_message': None, 'elapsed': 22, 'credit\_count': 1, 'notice': Non e, 'total\_count': 9580}, 'data': [{'id': 1, 'name': 'Bitcoin', 'symbo l': 'BTC', 'slug': 'bitcoin', 'num\_market\_pairs': 10392, 'date\_added': '2010-07-13T00:00:00.000Z', 'tags': ['mineable', 'pow', 'sha-256', 'st ore-of-value', 'state-channel', 'coinbase-ventures-portfolio', 'threearrows-capital-portfolio', 'polychain-capital-portfolio', 'binance-lab s-portfolio', 'blockchain-capital-portfolio', 'boostvc-portfolio', 'cm s-holdings-portfolio', 'dcg-portfolio', 'dragonfly-capital-portfolio', 'electric-capital-portfolio', 'fabric-ventures-portfolio', 'frameworkventures-portfolio', 'galaxy-digital-portfolio', 'huobi-capital-portfo lio', 'alameda-research-portfolio', 'a16z-portfolio', '1confirmation-p ortfolio', 'winklevoss-capital-portfolio', 'usv-portfolio', 'placehold er-ventures-portfolio', 'pantera-capital-portfolio', 'multicoin-capita l-portfolio', 'paradigm-portfolio', 'bitcoin-ecosystem'], 'max\_suppl y': 21000000, 'circulating\_supply': 19460781, 'total\_supply': 1946078 1, 'infinite\_supply': False, 'platform': None, 'cmc\_rank': 1, 'self\_re ported\_circulating\_supply': None, 'self\_reported\_market\_cap': None, 't vl\_ratio': None, 'last\_updated': '2023-08-18T07:37:00.000Z', 'quote':

#### In [6]:

type(data)

#### Out[6]:

dict

# (3) API JSON format to Dataframe

# In [7]:

pd.json\_normalize(data['data'])

# Out[7]:

	id	name	symbol	slug	num_market_pairs	date_added	tags
0	1	Bitcoin	ВТС	bitcoin	10392	2010-07- 13T00:00:00.000Z	[mineable, pow, sha- 256, store-of- value, state
1	1027	Ethereum	ETH	ethereum	7236	2015-08- 07T00:00:00.000Z	[pos, smart- contracts, ethereum- ecosystem, coi
2	825	Tether USDt	USDT	tether	59628	2015-02- 25T00:00:00.000Z	[payments, stablecoin, asset- backed- stablecoin
3	1839	BNB	BNB	bnb	1571	2017-07- 25T00:00:00.000Z	[marketplace, centralized- exchange, payments,
4	52	XRP	XRP	хгр	1032	2013-08- 04T00:00:00.000Z	[medium-of- exchange, enterprise- solutions, arr
5	3408	USD Coin	USDC	usd-coin	13653	2018-10- 08T00:00:00.000Z	[medium-of- exchange, stablecoin, asset- backed
6	2010	Cardano	ADA	cardano	892	2017-10- 01T00:00:00.000Z	[dpos, pos, platform, research, smart- contract
7	5426	Solana	SOL	solana	498	2020-04- 10T00:00:00.000Z	[pos, platform, solana- ecosystem, cms- holdings
8	74	Dogecoin	DOGE	dogecoin	770	2013-12- 15T00:00:00.000Z	[mineable, pow, scrypt, medium-of- exchange, me
9	1958	TRON	TRX	tron	809	2017-09- 13T00:00:00.000Z	[media, payments, tron- ecosystem, sec- security

	id	name	symbol	slug	num_market_pairs	date_added	tags
10	6636	Polkadot	DOT	polkadot- new	554	2020-08- 19T00:00:00.000Z	[substrate, polkadot, binance- chain, polkadot
11	3890	Polygon	MATIC	polygon	809	2019-04- 28T00:00:00.000Z	[pos, platform, enterprise- solutions, zero-kno
12	4943	Dai	DAI	multi- collateral- dai	2506	2019-11- 22T00:00:00.000Z	[defi, stablecoin, asset- backed- stablecoin, et
13	5994	Shiba Inu	SHIB	shiba-inu	606	2020-08- 01T00:00:00.000Z	[memes, ethereum- ecosystem, doggone- doggerel]
14	2	Litecoin	LTC	litecoin	970	2013-04- 28T00:00:00.000Z	[mineable, pow, scrypt, medium-of- exchange]

15 rows × 36 columns

# In [8]:

```
pd.set_option('display.max_columns', None)
pd.set_option('display.max_rows', None)
```

# In [9]:

```
df = pd.json_normalize(data['data'])
df['timestamp'] = pd.to_datetime('now')
df.head()
```

# Out[9]:

	id	name	symbol	slug	num_market_pairs	date_added	tags
0	1	Bitcoin	ВТС	bitcoin	10392	2010-07- 13T00:00:00.000Z	[mineable, pow, sha- 256, store-of- value, state
1	1027	Ethereum	ETH	ethereum	7236	2015-08- 07T00:00:00.000Z	[pos, smart- contracts, ethereum- ecosystem, coi
2	825	Tether USDt	USDT	tether	59628	2015-02- 25T00:00:00.000Z	[payments, stablecoin, asset- backed- stablecoin
3	1839	BNB	BNB	bnb	1571	2017-07- 25T00:00:00.000Z	[marketplace, centralized- exchange, payments,
4	52	XRP	XRP	хгр	1032	2013-08- 04T00:00:00.000Z	[medium-of- exchange, enterprise- solutions, arr
4							•

## (4) Automating the Data Pull

#### In [10]:

```
def api_runner():
    global df
    url = 'https://pro-api.coinmarketcap.com/v1/cryptocurrency/listings/latest'
    parameters = {
       'start':'1',
       'limit':'15'
       'convert': 'USD'
    headers = {
       'Accepts': 'application/json',
       'X-CMC_PRO_API_KEY': '0d11cefa-2730-441a-b86e-d5bc32d407d7',
    session = Session()
    session.headers.update(headers)
    try:
      response = session.get(url, params=parameters)
      data = json.loads(response.text)
      #print(data)
    except (ConnectionError, Timeout, TooManyRedirects) as e:
      print(e)
    # Use this if you just want to keep it in a dataframe
    df2 = pd.json_normalize(data['data'])
    df2['timestamp'] = pd.to_datetime('now')
    df = df.append(df2)
    # Use this if you want to create a csv and append data to it
    df = pd.json normalize(data['data'])
    df['timestamp'] = pd.to datetime('now')
    if not os.path.isfile(r'D:\IITG\portfolio_projects\api_automation\API.csv'):
        df.to csv(r'D:\IITG\portfolio projects\api automation\API.csv', header='column
    else:
        df.to csv(r'D:\IITG\portfolio projects\api automation\API.csv', mode='a', head
```

```
In [11]:
```

```
for i in range(333):
     api_runner()
     print('API Runner completed')
     sleep(60)
 exit()
API Runner completed
KeyboardInterrupt
                                           Traceback (most recent call las
t)
<ipython-input-11-d6baece82c3b> in <module>
      2
            api_runner()
            print('API Runner completed')
      3
            sleep(60)
---> 4
      5 exit()
KeyboardInterrupt:
In [22]:
```

```
# Use this to read in the data from the created csv
# df72 = pd.read_csv(r'D:\IITG\portfolio_projects\api_automation\API.csv')
# df72
```

#### In [12]:

Out[12]:

df

	id	name	symbol	slug	num_market_pairs	date_added	tags	max_supply
			- J				9-	
	1	Bitcoin	BTC	bitcoin	10392	2010-07-	[mineable,	2.100000e+
0						13T00:00:00.000Z	pow, sha-	
·							256, store-of-	
							value, state	
	1027	Ethereum	ETH	ethereum	7236	2015-08-	[pos, smart-	N
						07T00:00:00.000Z	contracts,	
1							ethereum-	
							ecosystem,	
							coi	
	825	Tether	USDT	tether	59628	2015-02-	[payments,	N
		USDt				25T00:00:00.000Z	stablecoin,	
2							asset-	
4								•

# (5) Transforming the Data

## In [13]:

```
pd.set_option('display.float_format', lambda x: '%.5f' % x)
df.head()
```

## Out[13]:

	id	name	symbol	slug	num_market_pairs	date_added	tags
0	1	Bitcoin	ВТС	bitcoin	10392	2010-07- 13T00:00:00.000Z	[mineable, pow, sha- 256, store-of- value, state
1	1027	Ethereum	ETH	ethereum	7236	2015-08- 07T00:00:00.000Z	[pos, smart- contracts, ethereum- ecosystem, coi
2	825	Tether USDt	USDT	tether	59628	2015-02- 25T00:00:00.000Z	[payments, stablecoin, asset- backed- stablecoin
3	1839	BNB	BNB	bnb	1571	2017-07- 25T00:00:00.000Z	[marketplace, centralized- exchange, payments,
4	52	XRP	XRP	хгр	1032	2013-08- 04T00:00:00.000Z	[medium-of- exchange, enterprise- solutions, arr
4							•

## In [14]:

```
df3 = df.groupby('name', sort=False)[['quote.USD.percent_change_1h','quote.USD.percent
df3
```

# Out[14]:

	quote.USD.percent_change_1h	quote.USD.percent_change_24h	quote.USD.perc
name			
Bitcoin	-0.40364	-7.45195	
Ethereum	-0.30895	-6.05881	
Tether USDt	-0.00984	0.10810	
BNB	-0.05823	-5.53148	
XRP	0.13087	-13.69195	
USD Coin	-0.00073	0.01151	
Cardano	-0.21501	-4.56540	
Solana	-0.34285	-5.03644	
Dogecoin	0.01707	-8.26741	
TRON	0.26655	-2.41224	
Polkadot	0.41150	-5.08404	
Polygon	-0.18483	-6.23230	
Dai	-0.10688	0.00342	
Shiba Inu	0.62967	-8.73826	
Litecoin	-0.19141	-13.38706	
4			•

# In [15]:

df4 = df3.stack()
df4

Out[15]:

name			
Bitcoin	<pre>quote.USD.percent_change_1h</pre>	-0.40364	
	quote.USD.percent_change_24h	-7.45195	
	quote.USD.percent_change_7d	-9.98055	
	quote.USD.percent_change_30d	-11.93245	
	quote.USD.percent_change_60d	0.17689	
	quote.USD.percent_change_90d	-1.48900	
Ethereum	quote.USD.percent_change_1h	-0.30895	
	quote.USD.percent_change_24h	-6.05881	
	quote.USD.percent_change_7d	-8.56953	
	quote.USD.percent_change_30d	-11.58795	
	quote.USD.percent change 60d	-2.02177	
	quote.USD.percent_change_90d	-6.74963	
Tether USDt	quote.USD.percent_change_1h	-0.00984	
recilei osbe	quote.USD.percent_change_24h	0.10810	
	quote.USD.percent_change_7d	0.12654	
	quote.USD.percent_change_30d	-0.04080	
	quote.USD.percent_change_60d	0.01238	
DND	quote.USD.percent_change_90d	-0.05063	
BNB	quote.USD.percent_change_1h	-0.05823	
	quote.USD.percent_change_24h	-5.53148	
	quote.USD.percent_change_7d	-9.03999	
	quote.USD.percent_change_30d	-10.01122	
	quote.USD.percent_change_60d	-10.19917	
	quote.USD.percent_change_90d	-29.07819	
XRP	quote.USD.percent_change_1h	0.13087	
	quote.USD.percent_change_24h	-13.69195	
	quote.USD.percent_change_7d	-19.71165	
	quote.USD.percent_change_30d	-35.68696	
	quote.USD.percent_change_60d	3.39751	
	quote.USD.percent_change_90d	9.29168	
USD Coin	quote.USD.percent_change_1h	-0.00073	
	quote.USD.percent_change_24h	0.01151	
In [16].	quote.USD.percent_change_7d	0.00512	
In [16]:	<pre>quote.USD.percent_change_30d</pre>	0.00610	
type(df3)	<pre>quote.USD.percent_change_60d</pre>	-0.00082	
cype(u.s)	quote.USD.percent_change_90d	0.00787	
Cardano Out[16]:	quote.USD.percent_change_1h	-0.21501	
	quote.USD.percent change 24h	-4.56540	
nandas cono	fAUMteDUSDFPSmeent_change_7d	-11.00401	
panuas.core.	quote.USD.percent_change_30d	-16.84636	
	quote.USD.percent_change_60d	1.09312	
In [17]:	quote.USD.percent_change_90d	-27.77821	
	quote.USD.percent_change_1h	-0.34285	
Solana type(df4)	quote.USD.percent_change_24h	-5.03644	
	quote.USD.percent_change_7d	-10.00891	
Out[17]:	quote.USD.percent_change_30d	-16.31701	
	quote USD nercent change 60d	42.83651	
pandas.core.	guote.USD.percent_change_60d series.Series quote.USD.percent_change_90d	9.32966	
Dogecoin	quote.USD.percent_change_1h	9.32900 0.01707	
PORECOTII	· · · · · · · · · · · · · · · · · · ·	-8.26741	
	quote.USD.percent_change_24h	-8.26741 -17.69884	
	quoto IICD noncont characted	- 1 / <b>.</b> n9884	
	quote.USD.percent_change_7d		
	quote.USD.percent_change_30d	-10.19283	
	<pre>quote.USD.percent_change_30d quote.USD.percent_change_60d</pre>	-10.19283 0.36942	
	quote.USD.percent_change_30d quote.USD.percent_change_60d quote.USD.percent_change_90d	-10.19283 0.36942 -15.23315	
TRON	quote.USD.percent_change_30d quote.USD.percent_change_60d quote.USD.percent_change_90d quote.USD.percent_change_1h	-10.19283 0.36942 -15.23315 0.26655	
TRON	quote.USD.percent_change_30d quote.USD.percent_change_60d quote.USD.percent_change_90d quote.USD.percent_change_1h quote.USD.percent_change_24h	-10.19283 0.36942 -15.23315 0.26655 -2.41224	
TRON	quote.USD.percent_change_30d quote.USD.percent_change_60d quote.USD.percent_change_90d quote.USD.percent_change_1h quote.USD.percent_change_24h quote.USD.percent_change_7d	-10.19283 0.36942 -15.23315 0.26655 -2.41224 -5.58609	
TRON	quote.USD.percent_change_30d quote.USD.percent_change_60d quote.USD.percent_change_90d quote.USD.percent_change_1h quote.USD.percent_change_24h quote.USD.percent_change_7d quote.USD.percent_change_30d	-10.19283 0.36942 -15.23315 0.26655 -2.41224 -5.58609 -8.80894	
TRON	quote.USD.percent_change_30d quote.USD.percent_change_60d quote.USD.percent_change_90d quote.USD.percent_change_1h quote.USD.percent_change_24h quote.USD.percent_change_7d	-10.19283 0.36942 -15.23315 0.26655 -2.41224 -5.58609	

```
Polkadot
              quote.USD.percent_change_1h
                                                    0.41150
In [18]:
               quote.USD.percent_change_24h
                                                   -5.08404
 df5 = df4.to frame(name= values)

df5 quote.USD.percent_change_30d
                                                   -9.93047
                                                  -14.03743
  df5
              quote.USD.percent_change_60d
                                                   -0.93471
               quote.USD.percent_change_90d
                                                  -16.18176
PUIVION:
               quote.USD.percent_change_1h
                                                   -0.18483
               quote.USD.percent_change_24h
                                                   -6.23230
                                                  -15.19025
                                         values
                                                  -22.28879
                                                   -2.99917
    name
                                                  -32.90834
    Bitcoin
             quote.USD.percent change 1h
                                          -0.40364
                                                   -0.10688
                                                    0.00342
            quote.USD.percent_change_24h
                                          -7.45195
                                                    0.02969
             quote.USD.percent_change_7d
                                                   -0.00472
                                          -9.98055
                                                   -0.00185
            quote.USD.percent_change_30d
                                         -11.93245
                                                   -0.00738
                                                    0.62967
            quote.USD.percent change 60d
                                          0.17689
                                                   -8.73826
            quote.USD.percent_change_90d
                                          -1.48900 -16.70246
                                                    7.74143
  Ethereum
             quote.USD.percent_change_1h
                                          -0.30895
                                                   16.48531
                                                   -3.07186
            quote.USD.percent_change_24h
                                          -6.05881
                                                   -0.19141
              quote.USD.percent_change_24h
                                                  -13.38706
               quote.USD.percent change 7d
                                                  -21.41339
In [19]:
              quote.USD.percent_change_30d
                                                  -30.13591
 type(df5)
              quote.USD.percent_change_60d
                                                  -15.65292
              quote.USD.percent_change_90d
                                                  -28.20093
```

dtypegjfloat64

pandas.core.frame.DataFrame

```
In [20]:
```

```
df5.count()
```

#### Out[20]:

values 90 dtype: int64

#### In [21]:

```
index = pd.Index(range(90))
```

## In [22]:

```
df6 = df5.reset_index()
df6
```

## Out[22]:

	name	level_1	values
0	Bitcoin	quote.USD.percent_change_1h	-0.40364
1	Bitcoin	quote.USD.percent_change_24h	-7.45195
2	Bitcoin	quote.USD.percent_change_7d	-9.98055
3	Bitcoin	quote.USD.percent_change_30d	-11.93245
4	Bitcoin	quote.USD.percent_change_60d	0.17689
5	Bitcoin	quote.USD.percent_change_90d	-1.48900
6	Ethereum	quote.USD.percent_change_1h	-0.30895
7	Ethereum	quote.USD.percent_change_24h	-6.05881
8	Ethereum	quote.USD.percent_change_7d	-8.56953

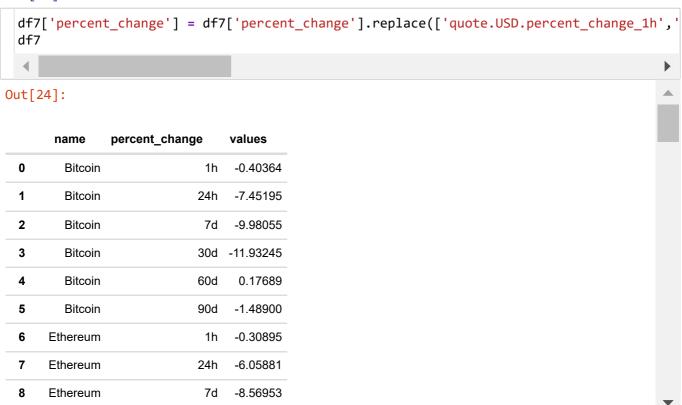
# In [23]:

```
df7 = df6.rename(columns={'level_1':'percent_change'})
df7
```

## Out[23]:

	name	percent_change	values
0	Bitcoin	quote.USD.percent_change_1h	-0.40364
1	Bitcoin	quote.USD.percent_change_24h	-7.45195
2	Bitcoin	quote.USD.percent_change_7d	-9.98055
3	Bitcoin	quote.USD.percent_change_30d	-11.93245
4	Bitcoin	quote.USD.percent_change_60d	0.17689
5	Bitcoin	quote.USD.percent_change_90d	-1.48900
6	Ethereum	quote.USD.percent_change_1h	-0.30895
7	Ethereum	quote.USD.percent_change_24h	-6.05881
8	Ethereum	quote.USD.percent_change_7d	-8.56953

#### In [24]:



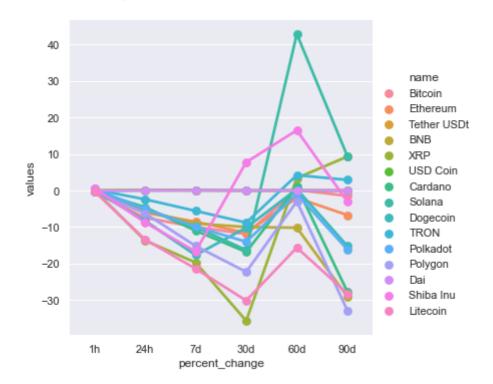
# (6) Visualizing the Data

#### In [25]:

```
sns.catplot(x='percent_change', y='values', hue='name', data=df7, kind='point')
```

### Out[25]:

<seaborn.axisgrid.FacetGrid at 0x2145e8536a0>



### In [26]:

```
df8 = df[['name', 'quote.USD.price', 'timestamp']]
df8
```

## Out[26]:

	name	quote.USD.price	timestamp
0	Bitcoin	26447.45162	2023-08-18 07:39:07.630322
1	Ethereum	1687.95924	2023-08-18 07:39:07.630322
2	Tether USDt	0.99975	2023-08-18 07:39:07.630322
3	BNB	218.44129	2023-08-18 07:39:07.630322
4	XRP	0.50797	2023-08-18 07:39:07.630322
5	USD Coin	1.00004	2023-08-18 07:39:07.630322
6	Cardano	0.26339	2023-08-18 07:39:07.630322
7	Solana	22.05945	2023-08-18 07:39:07.630322
8	Dogecoin	0.06223	2023-08-18 07:39:07.630322

# In [27]:

```
df9 = df8.query("name == 'Bitcoin'")
df9
```

## Out[27]:

	name	quote.USD.price	timestamp
0	Bitcoin	26447.45162	2023-08-18 07:39:07.630322
0	Bitcoin	26447.45162	2023-08-18 07:39:18.949811
0	Bitcoin	26443.69331	2023-08-18 07:40:20.355608
0	Bitcoin	26464.99743	2023-08-18 07:41:21.055115
0	Bitcoin	26464.66639	2023-08-18 07:42:22.258739
0	Bitcoin	26455.13874	2023-08-18 07:43:22.917310

### In [28]:

```
sns.set_theme(style='darkgrid')
sns.lineplot(x='timestamp', y='quote.USD.price', data=df9)
```

## Out[28]:

<AxesSubplot:xlabel='timestamp', ylabel='quote.USD.price'>

