### **Advanced Pandas**

# Bitcoin as a Hedge for Inflation — Is It Still a Good Option?

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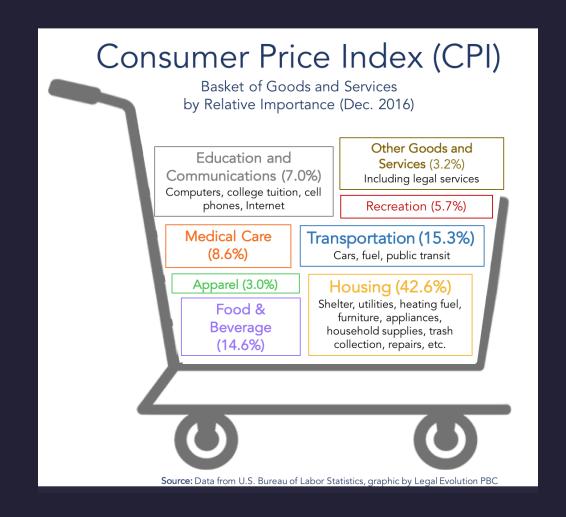
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https://www.nasdaq.com/articles/bitcoin-as-a-hedge-for-inflation-is-it-still-a-good-option#:~:text=Bitcoin has potential as an,add a level of risk.

#### Bitcoin as a hedge for inflation?

- Bitcoin price data in the coins table
- Get economic indicators for inflation
  - Consumer Price Index (CPI)
  - Global Price Index of all commodities (GPI)
- Compare bitcoin price with economic indicators

#### **Price Index**



#### O. Design database schema

#### 1. Extract data from a source

- i. identify endpoint, path, and query parameters from API documentation
- ii. request and get response from API

#### 2. Transform data

- i. select relevant values from response
- ii. transform data into a format that can be loaded into a database

#### 3. Load data into database

- i. create table
- ii. insert data into table

### **New tables**

Table: cpi

Column	Type		
date	VARCHAR(10)		
срі	FLOAT		

Table: gpi

Column	Type	
date	VARCHAR(10)	
gpi	FLOAT	

#### **Extract CPI and GPI from Alpha Vantage**

https://www.alphavantage.co/documentation/

- Endpoint
- Path
- Query parameters

#### Psuedocode

```
def indicators_etl(indicator: str)->None:
    """Extract, transform, and load economic indicators from Alpha Vantage API."""
    response = extract_indicators(indicator)
    data = transform_indicators(response)
    load_indicators(data)

indicators_etl("CPI")
```

#### **Extract**

```
def extract_indicators(indicator):
    url = "https://www.alphavantage.co/query"
    params = {
        "function": indicator,
        "apikey": yourkey
    }
    response = requests.get(url, params=params)

return response.json()
```

### Accessing items in a list/tuple/dict

list/tuple: [index]

• dict: [key]

- {"name": "Vancouver", "state": "BC", "country": "CA"}
- "Vancouver"
- "Montreal"

#### dict of list

```
cities = {
    "name": ["Montreal", "Toronto", "Vancouver", "Detroit"],
    "state": ["QC", "ON", "BC", "MI"],
    "country": ["CA", "CA", "US"]
}
```

- ["Montreal", "Toronto", "Vancouver", "Detroit"]
- "Montreal"

#### dict of dict

```
cities = {
    "Montreal": {"state": "QC", "country": "CA"},
    "Toronto": {"state": "ON", "country": "CA"},
    "Vancouver": {"state": "BC", "country": "CA"},
    "Detroit": {"state": "MI", "country": "US"}
}
```

- {"state": "QC", "country": "CA"}
- "QC"

```
cities = {
   "location": {
        "Montreal": {"state": "QC", "country": "CA"},
        "Toronto": {"state": "ON", "country": "CA"},
   "stats": {
       "Montreal":
            {"year": 2013, "population": 2000000, "area": 431.5},
            {"year": 2014, "population": 1980000, "area": 431.5}
        "Toronto": [
            {"year": 2013, "population": 2800000, "area": 630.2},
```

- {"year": 2013, "population": 2000000, "area": 431.5}
- "QC"
- 2000000

#### Select values from JSON response

```
"Meta Data": {
    "1. Information": "Daily Prices and Volumes for Digital Currency",
    "2. Digital Currency Code": "BTC",
"Time Series (Digital Currency Daily)": {
    "2023-10-30": {
        "4a. close (CNY)": "252122.24151800",
        "4b. close (USD)": "34456.58000000",
        . . .
    "2023-10-30": {...},
```

- "BTC"
- "34456.58000000"

#### Select CPI from response

```
"name": "Consumer Price Index for all Urban Consumers",
"interval": "monthly",
"unit": "index 1982-1984=100",
"data": [
        "date": "2023-09-01",
        "value": "307.789"
        "date": "2023-08-01",
        "value": "307.026"
```

#### **Transform**

```
def transform_indicators(response):
    data = []
    for values in response["data"]:
        record = (values["date"], values["value"])
        data.append(record)
    return data
```

#### Load

```
def load_data(conn, table, data):
    for record in data:
        query = f"INSERT INTO {table} VALUES {record}"
        conn.execute(query)
        conn.commit()
```

#### Read Bitcoin price, CPI, and GPI from database

```
coins_df = pd.read_sql("select * from coins order by date", conn)
cpi_df = pd.read_sql("select * from cpi order by date", conn)
gpi_df = pd.read_sql("select * from gpi order by date", conn)
```

#### Aggregate Bitcoin price by month

```
# First seven characters of the date is year-month (YYYY-MM-dd)
coins_df["month"] = coins_df["date"].str[:7]

# Group by month and get the last closing price of each month
coins_monthly_df = coins_df.groupby("month").agg({"close": "last"})
```

#### More aggregate functions

- last (first): last (first) value in a group
- nth: nth value in a group
- diff: difference from the previous value
- pct\_change : percentage change from the previous value
- nunique : number of unique values in a group

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https://pandas.pydata.org/pandas-docs/stable/reference/groupby.html#aggregation

### Calculate Bitcoin monthly return

```
coins_monthly_df["return"] = coins_monthly_df.agg({"close": "pct_change"})
```

#### Calculate monthly inflation rate (CPI)

```
# Month from date (YYYY-MM-dd)
cpi_df["month"] = cpi_df["date"].str[:7]

# get inflation rate
cpi_df["cpi_change"] = cpi_df.agg({"cpi": "pct_change"})
```

## **GPI** data

	date	gpi	
0	1992-01-01	•	
1	1992-02-01	•	
2	1992-03-01	•	
3	1992-04-01	•	
4	1992-05-01	•	
•••	•••	•••	
376	2023-05-01	157.134002	

## Replace . with missing value

If the value of a row is . (dot), replace it with missing value

- None: Python's built-in missing value
- pd.NA : Pandas' missing value
- np.nan: Numpy's missing value

### Conditional column creation using apply

#### apply

- method that applies a function along an axis of the DataFrame.
- axis=0 applies function to each column (default)
- axis=1 applies function to each row

```
def replace_dot(value):
    if value == ".":
        return None
    else:
        return value

gpi_df["gpi"] = gpi_df["gpi"].apply(replace_dot)
```

## **Filtering**

- query(): SQL-like syntax
- loc[]: label-based
- iloc[]: position-based`

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#### query for filtering

```
df.query('age > 25')
df.query('age > 25 and house == "Gryffindor"')
df.query('age > 25') = df['age'] * 2 # error
```

#### loc for filtering and updating

```
df.loc[df['age'] > 25]
df.loc[(df['age'] > 25) & (df['house'] == 'Gryffindor')]
df.loc[df['age'] > 25, 'age'] = df['age'] * 2  # ok
```

### More on loc[index, column]

```
# select row 0
df.loc[0]
# error: no row named 'age'
df.loc['age']
# select column 'age' (all rows)
df.loc[:, 'age']
# select rows 0 to 3, columns 'age' and 'name'
df.loc[0:3, ['age', 'name']]
# select rows where age > 25, columns 'age' and 'name'
df.loc[df['age'] > 25, ['age', 'name']]
```

### Conditional column creation using loc

```
gpi_df.loc[gpi_df["gpi"] == ".", "gpi"] = None
```

### **Conditional column creation**

Create a new column positive that is True if the monthly return is positive, and False otherwise

- **Q1.** apply
  - Write a function is\_positive that takes a value and returns True if the value is positive, and False otherwise
  - Use apply to apply the function to the return column
  - Assign the result to a new column positive
- **Q2.** loc
  - Use loc to select rows where return is positive
  - Assign True to the positive column in the selected rows

## **GPI** data

	date	gpi	
0	1992-01-01	None	
1	1992-02-01	None	
2	1992-03-01	None	
3	1992-04-01	None	
4	1992-05-01	None	
•••	•••	•••	
376	2023-05-01	157.134002	

#### Handling missing data

- isna(): returns True if the value is missing, False otherwise
- notna(): returns True if the value is not missing, False otherwise
- dropna(): drop rows with missing data

```
# Count missing data in each column
gpi_df.isna().agg('sum')

# Count non-missing data in each column
gpi_df.notna().agg('sum')

# Drop rows with missing data and assign to gpi_df
gpi_df = gpi_df.dropna()

# Drop rows with missing data and assign to gpi_df
gpi_df = gpi_df.dropna(subset=["gpi"])
```

## Missing data imputation

Year	Firm ID	Stock Price	Revenue	Earnings	Total Assets
2015	XYZ	85.50	1000	120	5000
2016	XYZ	90.00	1050	None	5200
2017	XYZ	None	1075	125	None
2018	XYZ	None	1100	130	5400
2019	XYZ	80.25	1150	None	5600
2020	XYZ	100.00	None	140	5800

#### Missing data imputation

```
# Fill missing data with 0 and assign to gpi_df
gpi_df = gpi_df.fillna(0)

# Fill missing data with the previous value and assign to gpi_df
gpi_df = gpi_df.fillna(method="ffill")

# Fill missing data with the next value and assign to gpi_df
gpi_df = gpi_df.fillna(method="bfill")

# Interpolate missing data with a linear method and assign to gpi_df
gpi_df = gpi_df.interpolate(method="linear")
```

#### Calculate inflation rate (GPI)

```
# Month from date (YYYY-MM-dd)
gpi_df["month"] = gpi_df["date"].str[:7]

# get inflation rate
gpi_df["cpi_change"] = gpi_df.agg({"gpi": "pct_change"})
```

#### Merge bitcoin price and economic indicators

```
df = coins_monthly_df.merge(cpi_df, on="month").merge(gpi_df, on="month")
cols = ['month', 'return', 'cpi_change', 'gpi_change']
df = df[cols].dropna()
df.head()
```

#### **Chaining methods**

```
cols = ['month', 'return', 'cpi_change', 'gpi_change']
df = coins monthly df.merge(cpi df, on="month").merge(gpi df, on="month").dropna()[cols]
cols = ['month', 'return', 'cpi_change', 'gpi_change']
df = (
    coins_monthly_df  # coins_monthly_df
.merge(cpi_df, on="month")  # merge with cpi_df
    coins_monthly_df
     .merge(gpi_df, on="month") # merge with gpi_df
    .dropna()
                                # drop rows with missing data
     [cols]
                                     # select columns
```

#### **Data Visualization**

- Matplotlib
- Seaborn
- Bokeh
- Altair
- Plotly

### **Plotly Express Syntax**

```
import plotly.express as px
fig = px.scatter(df, x="age", y="height")
fig.show()
```

https://plotly.com/python/plotly-express/ https://plotly.com/python/px-arguments/

### Line plot

```
# line plot for monthly return
fig = px.line(df, x="month", y="return")
fig.show()

# line plot for monthly return and inflation rate
fig = px.line(df, x="month", y=["return", "cpi_change"])
fig.show()
```

## Moving average (rolling)

```
# calculate 3-month moving average
df["return_ma"] = df["return"].rolling(3).mean()

# line plot for monthly return and 3-month moving average
fig = px.line(df, x="month", y=["return", "return_ma"])
fig.show()
```

#### Heatmap for correlation

```
# correlation matrix
corr = df.corr()

# heatmap
fig = px.imshow(corr, color_continuous_scale="Redor")
fig.show()
```

color scale: https://plotly.com/python/builtin-colorscales/

### Scatter plot with regression line

```
fig = px.scatter(df, x="inflation", y="return", trendline="ols")
fig.show()
```

### Regression using statsmodels

```
import statsmodels.api as sm

X = df["cpi_change"]
y = df["return"]
X = sm.add_constant(X)
model = sm.OLS(y, X).fit()
model.summary()
```

https://www.statsmodels.org/stable/index.html