

# Create a Web Application for an ETF Analyzer

In this Challenge assignment, you'll build a financial database and web application by using SQL, Python, and the Voilà library to analyze the performance of a hypothetical fintech ETF.

Instructions:

Use this notebook to complete your analysis of a fintech ETF that consists of four stocks: GOST, GS, PYPL, and SQ. Each stock has its own table in the `etf.db` database, which the `Starter_Code` folder also contains.

Analyze the daily returns of the ETF stocks both individually and as a whole. Then deploy the visualizations to a web application by using the Voilà library.

The detailed instructions are divided into the following parts:

- Analyze a single asset in the ETF
- Optimize data access with Advanced SQL queries
- Analyze the ETF portfolio
- Deploy the notebook as a web application

## Analyze a Single Asset in the ETF

For this part of the assignment, you'll use SQL queries with Python, Pandas, and hvPlot to analyze the performance of a single asset from the ETF.

Complete the following steps:

1. Write a SQL `SELECT` statement by using an f-string that reads all the PYPL data from the database. Using the SQL `SELECT` statement, execute a query that reads the PYPL data from the database into a Pandas DataFrame.
2. Use the `head` and `tail` functions to review the first five and the last five rows of the DataFrame. Make a note of the beginning and end dates that are available from this dataset. You'll use this information to complete your analysis.
3. Using hvPlot, create an interactive visualization for the PYPL daily returns. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.
4. Using hvPlot, create an interactive visualization for the PYPL cumulative returns. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.

## Optimize Data Access with Advanced SQL Queries

For this part of the assignment, you'll continue to analyze a single asset (PYPL) from the ETF. You'll use advanced SQL queries to optimize the efficiency of accessing data from the database.

Complete the following steps:

1. Access the closing prices for PYPL that are greater than 200 by completing the following steps:
  - Write a SQL `SELECT` statement to select the dates where the PYPL closing price was higher than 200.0.
  - Using the SQL statement, read the data from the database into a Pandas DataFrame, and then review the resulting DataFrame.
  - Select the "time" and "close" columns for those dates where the closing price was higher than 200.0.
2. Find the top 10 daily returns for PYPL by completing the following steps:
  - Write a SQL statement to find the top 10 PYPL daily returns. Make sure to do the following:
    - Use `SELECT` to select only the "time" and "daily\_returns" columns.
    - Use `ORDER` to sort the results in descending order by the "daily\_returns" column.
    - Use `LIMIT` to limit the results to the top 10 daily return values.
  - Using the SQL statement, read the data from the database into a Pandas DataFrame, and then review the resulting DataFrame.

## Analyze the ETF Portfolio

For this part of the assignment, you'll build the entire ETF portfolio and then evaluate its performance. To do so, you'll build the ETF portfolio by using SQL joins to combine all the data for each asset.

Complete the following steps:

1. Write a SQL query to join each table in the portfolio into a single DataFrame. To do so, complete the following steps:
  - Use a SQL inner join to join each table on the "time" column. Access the "time" column in the `GDOT` table via the `GDOT.time` syntax. Access the "time" columns from the other tables via similar syntax.
  - Using the SQL query, read the data from the database into a Pandas DataFrame. Review the resulting DataFrame.
2. Create a DataFrame that averages the "daily\_returns" columns for all four assets. Review the resulting DataFrame

DataFrame.

**Hint** Assuming that this ETF contains equally weighted returns, you can average the returns for each asset to get the average returns of the portfolio. You can then use the average returns of the portfolio to calculate the annualized returns and the cumulative returns. For the calculation to get the average daily returns for the portfolio, use the following code:

```
etf_portfolio_returns = etf_portfolio['daily_returns'].mean(axis=1)
```

You can use the average daily returns of the portfolio the same way that you used the daily returns of a single asset.

3. Use the average daily returns in the `etf_portfolio_returns` DataFrame to calculate the annualized returns for the portfolio. Display the annualized return value of the ETF portfolio.

**Hint** To calculate the annualized returns, multiply the mean of the `etf_portfolio_returns` values by 252.

To convert the decimal values to percentages, multiply the results by 100.

1. Use the average daily returns in the `etf_portfolio_returns` DataFrame to calculate the cumulative returns of the ETF portfolio.
2. Using `hvPlot`, create an interactive line plot that visualizes the cumulative return values of the ETF portfolio. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.

## Deploy the Notebook as a Web Application

For this part of the assignment, complete the following steps:

1. Use the `Voilà` library to deploy your notebook as a web application. You can deploy the web application locally on your computer.
2. Take a screen recording or screenshots to show how the web application appears when using `Voilà`. Include the recording or screenshots in the `README.md` file for your GitHub repository.

Review the following code which imports the required libraries, initiates your SQLite database, populates the database with records from the `etf.db` seed file that was included in your `Starter_Code` folder, creates the database engine, and confirms that data tables that it now contains.

```
['GDOT', 'GS', 'PYPL', 'SQ']
```

# Analyze a single asset in the FinTech ETF

For this part of the assignment, you'll use SQL queries with Python, Pandas, and hvPlot to analyze the performance of a single asset from the ETF.

Complete the following steps:

1. Write a SQL `SELECT` statement by using an f-string that reads all the PYPL data from the database. Using the SQL `SELECT` statement, execute a query that reads the PYPL data from the database into a Pandas DataFrame.
2. Use the `head` and `tail` functions to review the first five and the last five rows of the DataFrame. Make a note of the beginning and end dates that are available from this dataset. You'll use this information to complete your analysis.
3. Using hvPlot, create an interactive visualization for the PYPL daily returns. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.
4. Using hvPlot, create an interactive visualization for the PYPL cumulative returns. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.

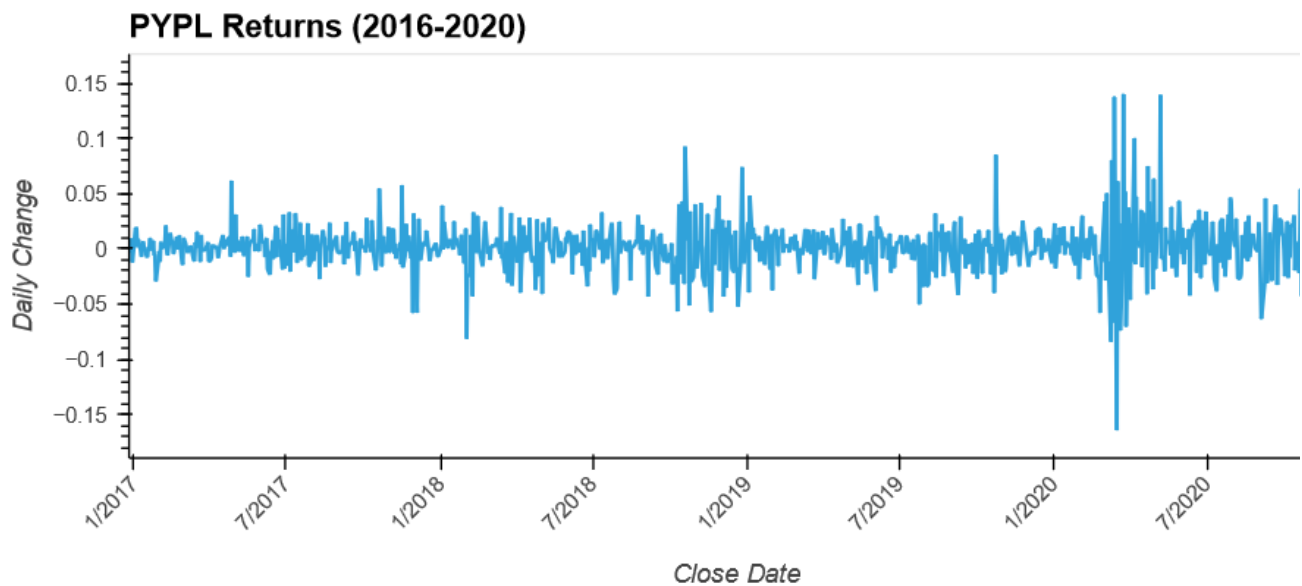
Step 1: Write a SQL `SELECT` statement by using an f-string that reads all the PYPL data from the database. Using the SQL `SELECT` statement, execute a query that reads the PYPL data from the database into a Pandas DataFrame.

Step 2: Use the `head` and `tail` functions to review the first five and the last five rows of the DataFrame. Make a note of the beginning and end dates that are available from this dataset. You'll use this information to complete your analysis.

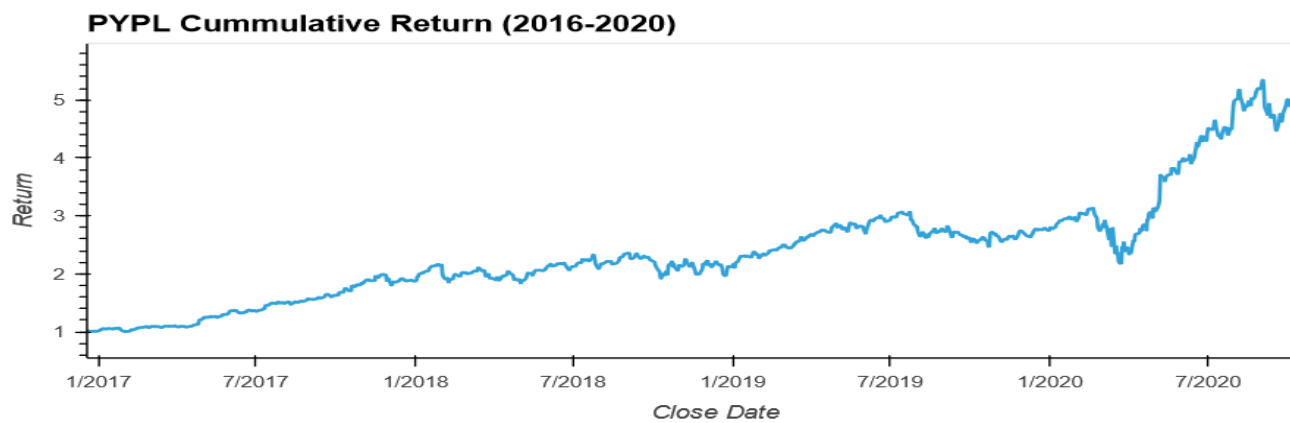
|   | time                       | open  | high  | low   | close | volume  | daily_returns |
|---|----------------------------|-------|-------|-------|-------|---------|---------------|
| 0 | 2016-12-16 00:00:00.000000 | 39.90 | 39.90 | 39.12 | 39.32 | 7298861 | -0.005564     |
| 1 | 2016-12-19 00:00:00.000000 | 39.40 | 39.80 | 39.11 | 39.45 | 3436478 | 0.003306      |
| 2 | 2016-12-20 00:00:00.000000 | 39.61 | 39.74 | 39.26 | 39.74 | 2940991 | 0.007351      |
| 3 | 2016-12-21 00:00:00.000000 | 39.84 | 40.74 | 39.82 | 40.09 | 5826704 | 0.008807      |
| 4 | 2016-12-22 00:00:00.000000 | 40.04 | 40.09 | 39.54 | 39.68 | 4338385 | -0.010227     |

|     | time                       | open   | high   | low      | close   | volume  | daily_returns |
|-----|----------------------------|--------|--------|----------|---------|---------|---------------|
| 994 | 2020-11-30 00:00:00.000000 | 212.51 | 215.83 | 207.0900 | 214.200 | 8992681 | 0.013629      |
| 995 | 2020-12-01 00:00:00.000000 | 217.15 | 220.57 | 214.3401 | 216.520 | 9148174 | 0.010831      |
| 996 | 2020-12-02 00:00:00.000000 | 215.60 | 215.75 | 210.5000 | 212.660 | 6414746 | -0.017827     |
| 997 | 2020-12-03 00:00:00.000000 | 213.33 | 216.93 | 213.1100 | 214.680 | 6463339 | 0.009499      |
| 998 | 2020-12-04 00:00:00.000000 | 214.88 | 217.28 | 213.0100 | 217.235 | 2118319 | 0.011901      |

Step 3: Using hvPlot, create an interactive visualization for the PYPL daily returns. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.



Step 4: Using hvPlot, create an interactive visualization for the PYPL cumulative returns. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.



# Optimize the SQL Queries

For this part of the assignment, you'll continue to analyze a single asset (PYPL) from the ETF. You'll use advanced SQL queries to optimize the efficiency of accessing data from the database.

Complete the following steps:

1. Access the closing prices for PYPL that are greater than 200 by completing the following steps:
2. Access the closing prices for PYPL that are greater than 200 by completing the following steps:
  - Write a SQL `SELECT` statement to select the dates where the PYPL closing price was higher than 200.0.
  - Select the "time" and "close" columns for those dates where the closing price was higher than 200.0.
  - Using the SQL statement, read the data from the database into a Pandas DataFrame, and then review the resulting DataFrame.
3. Find the top 10 daily returns for PYPL by completing the following steps:
  - Write a SQL statement to find the top 10 PYPL daily returns. Make sure to do the following:
    - Use `SELECT` to select only the "time" and "daily\_returns" columns.
    - Use `ORDER` to sort the results in descending order by the "daily\_returns" column.
    - Use `LIMIT` to limit the results to the top 10 daily return values.
  - Using the SQL statement, read the data from the database into a Pandas DataFrame, and then review the resulting DataFrame.

## Step 1: Access the closing prices for PYPL that are greater than 200 by completing the following steps:

- Write a SQL `SELECT` statement to select the dates where the PYPL closing price was higher than 200.0.
- Select the "time" and "close" columns for those dates where the closing price was higher than 200.0.
- Using the SQL statement, read the data from the database into a Pandas DataFrame, and then review the resulting DataFrame.

|   | time                       | close  |
|---|----------------------------|--------|
| 0 | 2020-08-05 00:00:00.000000 | 202.92 |
| 1 | 2020-08-06 00:00:00.000000 | 204.09 |
| 2 | 2020-08-25 00:00:00.000000 | 201.71 |
| 3 | 2020-08-26 00:00:00.000000 | 203.53 |
| 4 | 2020-08-27 00:00:00.000000 | 204.34 |
| 5 | 2020-08-28 00:00:00.000000 | 204.48 |
| 6 | 2020-08-31 00:00:00.000000 | 203.95 |
| 7 | 2020-09-01 00:00:00.000000 | 208.92 |
| 8 | 2020-09-02 00:00:00.000000 | 210.82 |
| 9 | 2020-09-03 00:00:00.000000 | 205.07 |

Step 2: Find the top 10 daily returns for PYPL by completing the following steps:

- Write a SQL statement to find the top 10 PYPL daily returns. Make sure to do the following:

- \* Use `SELECT` to select only the “time” and “daily\_returns” columns.

- \* Use `ORDER` to sort the results in descending order by the “daily\_returns” column.

- \* Use `LIMIT` to limit the results to the top 10 daily return values.

- Using the SQL statement, read the data from the database into a Pandas DataFrame, and then review the resulting DataFrame.



|          | time                       | daily_returns |
|----------|----------------------------|---------------|
| <b>0</b> | 2020-03-24 00:00:00.000000 | 0.140981      |
| <b>1</b> | 2020-05-07 00:00:00.000000 | 0.140318      |
| <b>2</b> | 2020-03-13 00:00:00.000000 | 0.138700      |
| <b>3</b> | 2020-04-06 00:00:00.000000 | 0.100877      |
| <b>4</b> | 2018-10-19 00:00:00.000000 | 0.093371      |
| <b>5</b> | 2019-10-24 00:00:00.000000 | 0.085912      |
| <b>6</b> | 2020-11-04 00:00:00.000000 | 0.080986      |
| <b>7</b> | 2020-03-10 00:00:00.000000 | 0.080863      |
| <b>8</b> | 2020-04-22 00:00:00.000000 | 0.075321      |
| <b>9</b> | 2018-12-26 00:00:00.000000 | 0.074656      |

# Analyze the Fintech ETF Portfolio

For this part of the assignment, you'll build the entire ETF portfolio and then evaluate its performance. To do so, you'll build the ETF portfolio by using SQL joins to combine all the data for each asset.

Complete the following steps:

1. Write a SQL query to join each table in the portfolio into a single DataFrame. To do so, complete the following steps:
  - Use a SQL inner join to join each table on the "time" column. Access the "time" column in the `GDOT` table via the `GDOT.time` syntax. Access the "time" columns from the other tables via similar syntax.
  - Using the SQL query, read the data from the database into a Pandas DataFrame. Review the resulting DataFrame.
2. Create a DataFrame that averages the "daily\_returns" columns for all four assets. Review the resulting DataFrame.

**Hint** Assuming that this ETF contains equally weighted returns, you can average the returns for each asset to get the average returns of the portfolio. You can then use the average returns of the portfolio to calculate the annualized returns and the cumulative returns. For the calculation to get the average daily returns for the portfolio, use the following code:

```
etf_portfolio_returns = etf_portfolio['daily_returns'].mean(axis=1)
```

You can use the average daily returns of the portfolio the same way that you used the daily returns of a single asset.

3. Use the average daily returns in the `etf_portfolio_returns` DataFrame to calculate the annualized returns for the portfolio. Display the annualized return value of the ETF portfolio.

**Hint** To calculate the annualized returns, multiply the mean of the `etf_portfolio_returns` values by 252.

To convert the decimal values to percentages, multiply the results by 100.

1. Use the average daily returns in the `etf_portfolio_returns` DataFrame to calculate the cumulative returns of the ETF portfolio.
2. Using `hvPlot`, create an interactive line plot that visualizes the cumulative return values of the ETF portfolio. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.

Step 1: Write a SQL query to join each table in the portfolio into a single DataFrame. To do so, complete the following steps:

- Use a SQL inner join to join each table on the “time” column. Access the “time” column in the `GDOT` table via the `GDOT.time` syntax. Access the “time” columns from the other tables via similar syntax.
- Using the SQL query, read the data from the database into a Pandas DataFrame. Review the resulting DataFrame.

|                            | daily_returns | daily_returns | daily_returns | daily_returns |
|----------------------------|---------------|---------------|---------------|---------------|
| time                       |               |               |               |               |
| 2016-12-16 00:00:00.000000 | -0.005564     | -0.023218     | -0.016708     | 0.017339      |
| 2016-12-19 00:00:00.000000 | 0.003306      | -0.007923     | 0.000795      | -0.001043     |
| 2016-12-20 00:00:00.000000 | 0.007351      | 0.001261      | 0.016602      | 0.009053      |
| 2016-12-21 00:00:00.000000 | 0.008807      | 0.001679      | -0.006911     | -0.007591     |
| 2016-12-22 00:00:00.000000 | -0.010227     | 0.006077      | -0.005178     | -0.023644     |
| ...                        | ...           | ...           | ...           | ...           |
| 2020-11-30 00:00:00.000000 | 0.013629      | -0.043750     | -0.021266     | -0.007153     |
| 2020-12-01 00:00:00.000000 | 0.010831      | 0.004482      | 0.006549      | -0.037823     |
| 2020-12-02 00:00:00.000000 | -0.017827     | -0.027328     | 0.024387      | -0.004384     |
| 2020-12-03 00:00:00.000000 | 0.009499      | 0.027523      | -0.008959     | 0.016921      |
| 2020-12-04 00:00:00.000000 | 0.011901      | 0.001860      | 0.012520      | 0.010151      |

999 rows × 4 columns

Step 2: Create a DataFrame that averages the “daily\_returns” columns for all four assets. Review the resulting DataFrame.

```
time
2016-12-16 00:00:00.000000    -0.007038
2016-12-19 00:00:00.000000    -0.001216
2016-12-20 00:00:00.000000     0.008567
2016-12-21 00:00:00.000000    -0.001004
2016-12-22 00:00:00.000000    -0.008243
...
2020-11-30 00:00:00.000000    -0.014635
2020-12-01 00:00:00.000000    -0.003990
2020-12-02 00:00:00.000000    -0.006288
2020-12-03 00:00:00.000000     0.011246
2020-12-04 00:00:00.000000     0.009108
Length: 999, dtype: float64
```

Step 3: Use the average daily returns in the `etf_portfolio_returns` DataFrame to calculate the annualized returns for the portfolio. Display the annualized return value of the ETF portfolio.

The Portfolio Annualized Return is 43.83%

Step 4: Use the average daily returns in the `etf_portfolio_returns` DataFrame to calculate the cumulative returns of the ETF portfolio.

The final cumulative return for the combined portfolio is 441.83%

Step 5: Using `hvPlot`, create an interactive line plot that visualizes the cumulative return values of the ETF portfolio. Reflect the "time" column of the DataFrame on the x-axis. Make sure that you professionally style and format your visualization to enhance its readability.

```
time
2016-12-16 00:00:00.000000    0.992962
2016-12-19 00:00:00.000000    0.991755
2016-12-20 00:00:00.000000    1.000251
2016-12-21 00:00:00.000000    0.999246
2016-12-22 00:00:00.000000    0.991010
...
2020-11-30 00:00:00.000000    4.374534
2020-12-01 00:00:00.000000    4.357078
2020-12-02 00:00:00.000000    4.329679
2020-12-03 00:00:00.000000    4.378371
2020-12-04 00:00:00.000000    4.418250
Length: 999, dtype: float64
```

