

Lesson 06: Selection and Filtering

Data Science with Python – BSc Course

Data Science Program

45 Minutes

After this lesson, you will be able to:

- Select columns using bracket and dot notation
- Access rows with iloc (position) and loc (label)
- Filter data using boolean conditions
- Combine multiple conditions with & and —

Finance Application: Screen stocks by price, volume, and other criteria.

Selection and filtering extract relevant data for analysis

Column Selection Methods

```
df['AAPL']
```

Single column (Series)

```
df[['AAPL', 'MSFT']]
```

Multiple columns (DataFrame)

```
df.AAPL
```

Attribute access (simple names)

```
df.loc[:, 'AAPL':'GOOGL']
```

Range of columns

iloc vs loc

iloc (Integer Location)

Position-based indexing

```
df.iloc[0]
```

```
df.iloc[0:5, 1:3]
```

Uses: 0, 1, 2, ...

loc (Label Location)

Label-based indexing

```
df.loc['2024-01-02']
```

```
df.loc[:, 'AAPL']
```

Uses: dates, names

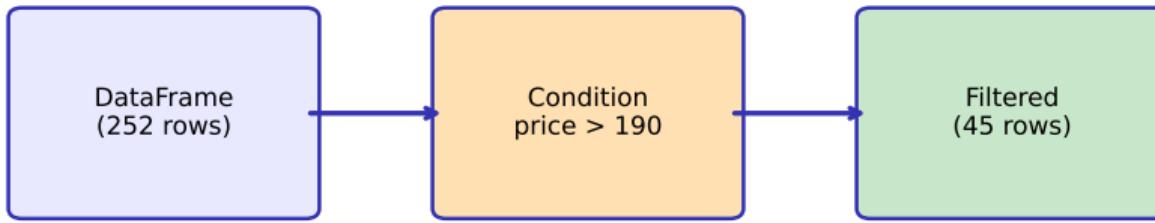
iloc: exclusive end | loc: inclusive end

Boolean Masking

| AAPL | df["AAPL"] > 188 | Mask | Result |
|------|------------------|-------|--------|
| 185 | | False | 190 |
| 190 | | True | 195 |
| 188 | → | False | → |
| 195 | | True | |
| 182 | | False | |

Boolean mask filters rows where condition is True

Conditional Filtering Flow



```
df_filtered = df[df["AAPL"] > 190]
```

Multiple Conditions

AND: &

```
(df["AAPL"] > 185) &  
(df["MSFT"] > 380)
```

OR: |

```
(df["AAPL"] > 200) |  
(df["MSFT"] > 400)
```

NOT: ~

```
~(df["AAPL"] > 190)
```

Always use parentheses around each condition!

Chained Filtering with query()

Traditional

```
df[(df["AAPL"] > 185) &  
    (df["Volume"] > 1e6)]
```

query() Method

```
df.query("AAPL > 185 and  
         Volume > 1e6")
```

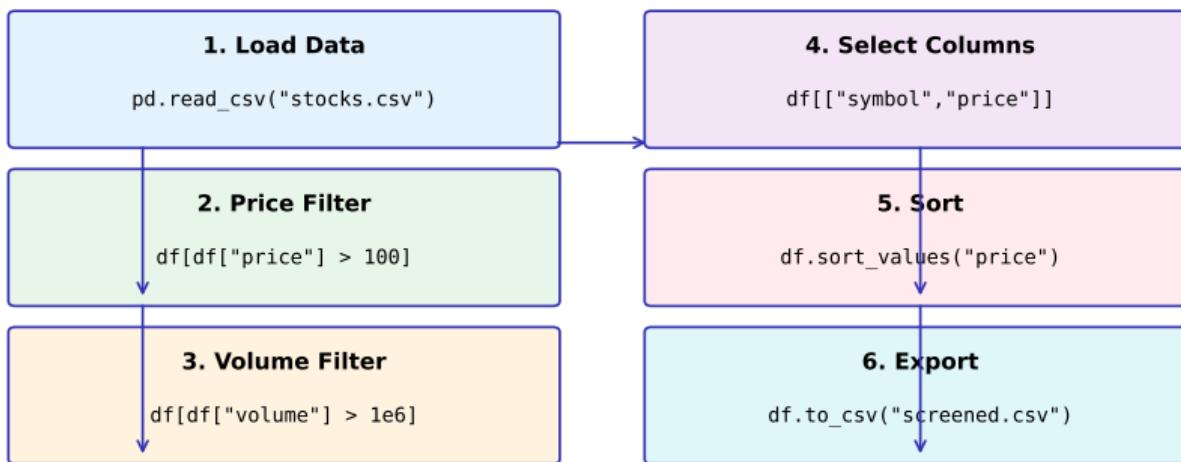
query() is more readable for complex filters

Membership: isin()

```
df[df["Symbol"].isin(["AAPL", "MSFT", "GOOGL"])]
```

Selection Methods Comparison

| Method | Use Case | Returns |
|-----------------------------------|------------------|-----------|
| <code>df["col"]</code> | Single column | Series |
| <code>df[["col1", "col2"]]</code> | Multiple columns | DataFrame |
| <code>df.iloc[0]</code> | Row by position | Series |
| <code>df.loc["date"]</code> | Row by label | Series |
| <code>df[df.col > x]</code> | Filter rows | DataFrame |



Combine filters to build powerful stock screeners

Hands-on Exercise (25 min)

Build a stock screener:

- ① Load stock data from CSV
- ② Select only AAPL and MSFT columns
- ③ Filter rows where AAPL > 185
- ④ Filter rows where AAPL > 185 AND MSFT > 375
- ⑤ Use query() for the same filter
- ⑥ Select first 10 trading days using iloc
- ⑦ Sort by AAPL price descending

Stock screeners are fundamental tools in finance

Lesson Summary

Key Takeaways:

- `df["col"]` selects single column as Series
- `iloc` uses integer positions; `loc` uses labels
- Boolean conditions create True/False masks
- Combine conditions with `&` (and) and `—` (or)
- `query()` is cleaner for complex filters

Next Lesson: Missing Data and Cleaning

Week 1 complete! You can now load, explore, and filter data