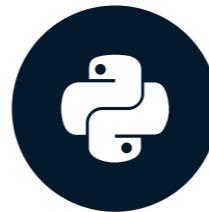


Introduction to ETL and ELT Pipelines

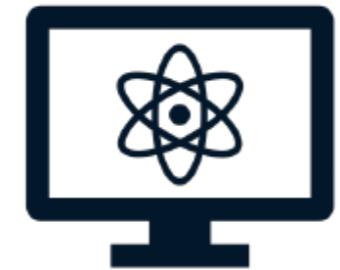
ETL AND ELT IN PYTHON



Jake Roach
Data Engineer



Business Intelligence



Machine Learning



Artificial Intelligence

Data pipelines

... are responsible for moving data from a source to a destination, and transforming it somewhere along the way.

Source → Destination



API



PostgreSQL



ETL

- Extract, transform, load
- Traditional data pipeline design pattern
- Sources may be tabular or non-tabular
- Leverage Python with `pandas`

ELT

- Extract, load, transform
- More recent pattern
- Data warehouses
- Typically tabular data

Extract, transform, load (ETL)

```
def load(data_frame, target_table):
    # Some custom-built Python logic to load data to SQL
    data_frame.to_sql(name=target_table, con=POSTGRES_CONNECTION)
    print(f"Loading data to the {target_table} table")

# Now, run the data pipeline
extracted_data = extract(file_name="raw_data.csv")
transformed_data = transform(data_frame=extracted_data)
load(data_frame=transformed_data, target_table="cleaned_data")
```

Extracting data from raw_data.csv

Transforming data to remove 'null' records

Loading data to the cleaned_data table

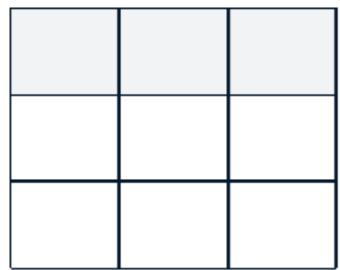
Extract, load, transform (ELT)

...

```
def transform(source_table, target_table):
    data_warehouse.run_sql("""
        CREATE TABLE {target_table} AS
        SELECT
            <field-name>, <field-name>, ...
        FROM {source_table};
    """)
```

```
# Similar to ETL pipelines, call the extract, load, and transform functions
extracted_data = extract(file_name="raw_data.csv")
load(data_frame=extracted_data, table_name="raw_data")
transform(source_table="raw_data", target_table="cleaned_data")
```

We'll also take a look at...



Tabular and
non-tabular data

- Data transformation
- Writing data to disk,
SQL databases

- Unit-testing
- Monitoring
- Deploying to production



Let's practice!

ETL AND ELT IN PYTHON

Building ETL and ELT Pipelines

ETL AND ELT IN PYTHON



Jake Roach
Data Engineer

Extract Data from a CSV File

```
import pandas as pd
```

```
# Read in the CSV file to a DataFrame  
data_frame = pd.read_csv("raw_data.csv")
```

```
# Output the first few rows  
data_frame.head()
```

	name	num_firms	total_income
0	Advertising	58	3892.41
1	Apparel	39	5422.69
...			
49	Trucking	35	17324.36

`read_csv()`

- Takes a file path, returns a DataFrame
- `delimiter`, `header`, `engine`

`.head()`

- Outputs the first `n` number of a DataFrame

Filtering a DataFrame

	name	num_firms	total_income
0	Advertising	58	3892.41
1	Apparel	39	5422.69
...			
49	Trucking	35	17324.36

	name	num_firms
1	Apparel	39
37	Apparel	61

```
# First, by rows  
  
data_frame.loc[data_frame["name"] == "Apparel", :]  
  
# Then, by columns  
  
data_frame.loc[:, ["name", "num_firms"]]
```

.loc

- Filters a DataFrame
- `:` means "all"

Write a DataFrame to a CSV File

```
# Write a DataFrame to a .csv file  
data_frame.to_csv("cleaned_data.csv")
```

.to_csv()

- Takes a `path`, creates DataFrame from file stored at that `path`
- Can take other parameters to customize the output

Other options, like:

.to_json() , .to_excel() , .to_sql()

Running SQL Queries

```
data_warehouse.execute( # Use Python clients or other tools to run SQL queries
    """
CREATE TABLE total_sales AS
    SELECT
        ds,
        SUM(sales)
    FROM raw_sales_data
    GROUP BY ds;
    """
)

```

- Tools like `.execute()` to run SQL queries

Putting it all together!

```
# Define extract(), transform(), and load() functions
...
def transform(data_frame, value):
    return data_frame.loc[data_frame["name"] == value, ["name", "num_firms"]]

# First, extract data from a .csv
extracted_data = extract(file_name="raw_data.csv")

# Then, transform the `extracted_data`
transformed_data = transform(data_frame=extracted_data, value="Apparel")

# Finally, load the `transformed_data`
load(data_frame=transformed_data, file_name="cleaned_data.csv")
```

Let's practice!

ETL AND ELT IN PYTHON