Simple Weather ETL (Step-by-Step, with Code)

A copy-pasteable, beginner-friendly guide to build a **non-automated** ETL pipeline that pulls **current weather** from OpenWeather, stores raw JSON locally, transforms it to a tidy table, and loads it into **MySQL** (plus an optional CSV fallback). You can run the pipeline manually from your terminal—no schedulers.

What you'll build

- Extract: Call OpenWeather's Current Weather endpoint for a list of cities.
- **Transform**: Normalize JSON → a clean tabular schema.
- Load: Insert rows into a MySQL table (or write to CSV if you don't have MySQL yet).
- **Artifacts**: Raw JSON in data/raw/..., processed CSV/Parquet in data/processed/...

0) Prerequisites (10-20 min)

- Python 3.9+
- MySQL Server (8.x recommended) + MySQL Workbench (optional)
- An OpenWeather API key (free): create an account → User dashboard → API keys. We'll use the Current Weather Data endpoint.

If you can't use MySQL yet, you can still complete this guide using the **CSV fallback** and add MySQL later.

1) Project Structure

```
weather-etl/
─ README.md

    ⊢ requirements.txt

    ─ config/

  ├ config.sample.yaml
  └ config.yaml
                          # your real config (not committed)
├ data/
  ⊢ raw/
                          # raw JSON dumps
  └ processed/
                          # cleaned CSV/Parquet
⊢ etl/
  ⊢ extract.py
  ⊢ transform.py
   ├ load.py
  └ pipeline.py
```

```
└─ sql/
└─ create_tables.sql
```

Create these folders/files as you go. We'll fill them in.

2) Create & Activate a Virtual Environment

```
# macOS/Linux
python3 -m venv .venv
source .venv/bin/activate

# Windows (PowerShell)
python -m venv .venv
.venv\Scripts\Activate.ps1
```

3) Dependencies

requirements.txt

```
requests
pandas
PyYAML
SQLAlchemy
PyMySQL
python-dateutil
pyarrow # for Parquet (optional but nice)
```

Install:

```
pip install -r requirements.txt
```

4) Git Hygiene

.gitignore

```
# venv
.venv/
```

```
# Python
__pycache__/
*.pyc

# Data & secrets (never commit!)
config/config.yaml
data/
logs/
```

5) Configuration

config/config.sample.yaml (commit this)

```
openweather:
  api_base: "https://api.openweathermap.org/data/2.5/weather"
  api_key: "REPLACE_ME"
  units: "metric"
                         # metric | imperial | standard
run:
  cities: ["Kathmandu,NP", "Pokhara,NP"] # city, country_code
  save_parquet: true
mysql:
  enabled: true
  host: "localhost"
  port: 3306
  user: "weather_user"
  password: "REPLACE_ME"
  database: "weather"
  table: "weather_observations"
```

Make a copy named config.yaml and put your real values there. Don't commit it.

6) MySQL Setup (5-10 min)

In MySQL Workbench or CLI, create a database, a user, and the table.

Create DB & user

```
CREATE DATABASE IF NOT EXISTS weather CHARACTER SET utf8mb4 COLLATE utf8mb4_0900_ai_ci;
CREATE USER IF NOT EXISTS 'weather_user'@'%' IDENTIFIED BY
'YOUR_STRONG_PASSWORD';
GRANT ALL PRIVILEGES ON weather.* TO 'weather_user'@'%';
FLUSH PRIVILEGES;
```

Table schema (sql/create_tables.sql)

```
CREATE TABLE IF NOT EXISTS weather_observations (
 id BIGINT PRIMARY KEY AUTO_INCREMENT,
 city_id BIGINT NULL,
 city_name VARCHAR(128) NOT NULL,
 country_code VARCHAR(8) NULL,
 lat DECIMAL(9,6) NULL,
 lon DECIMAL(9,6) NULL,
 observation_time_utc DATETIME NOT NULL,
 temp_c DECIMAL(6,2) NULL,
 feels_like_c DECIMAL(6,2) NULL,
 temp_min_c DECIMAL(6,2) NULL,
 temp_max_c DECIMAL(6,2) NULL,
 pressure_hpa INT NULL,
 humidity_pct INT NULL,
 wind_speed_ms DECIMAL(6,2) NULL,
 wind_deg INT NULL,
 clouds_pct INT NULL,
 weather_main VARCHAR(64) NULL,
 weather_desc VARCHAR(256) NULL,
 rain_1h_mm DECIMAL(6,2) NULL,
 snow_1h_mm DECIMAL(6,2) NULL,
 source VARCHAR(64) NOT NULL,
 ingested_at_utc DATETIME NOT NULL,
 UNIQUE KEY uq_city_time (city_id, observation_time_utc)
) ENGINE=InnoDB;
```

Run it once:

```
# if you have mysql client
mysql -u weather_user -p -h 127.0.0.1 weather < sql/create_tables.sql</pre>
```

Note: We keep both an AUTO_INCREMENT id and a unique key on (city_id, observation_time_utc) to prevent duplicates when you re-run the same fetch.

7) Extract

etl/extract.py

```
from __future__ import annotations
import json
from pathlib import Path
from typing import Dict, List, Any
from datetime import datetime, timezone
import time
import requests
RAW_DIR = Path("data/raw")
def _timestamp() -> str:
    return datetime.now(timezone.utc).strftime("%Y%m%dT%H%M%SZ")
def fetch_current_weather_city(api_base: str, api_key: str, city: str, units:
str = "metric") -> Dict[str, Any]:
    """Fetch current weather for a single city using q=city,country.
   Returns parsed JSON (dict). Raises for non-200.
   params = {"q": city, "appid": api_key, "units": units}
    resp = requests.get(api_base, params=params, timeout=20)
    resp.raise_for_status()
   return resp.json()
def extract_current_weather(api_base: str, api_key: str, cities: List[str],
units: str = "metric", pause_secs: float = 1.0) -> List[Dict[str, Any]]:
    """Fetch multiple cities with a small pause to be polite to the API.
   Saves each raw JSON to data/raw/YYYYMMDD/<city>_<timestamp>.json
   Returns the list of JSON dicts in memory as well.
   day_dir = RAW_DIR / datetime.now(timezone.utc).strftime("%Y%m%d")
   day_dir.mkdir(parents=True, exist_ok=True)
    results: List[Dict[str, Any]] = []
    for city in cities:
        data = fetch_current_weather_city(api_base, api_key, city, units)
        # persist raw JSON
        city_sanitized = city.replace(",", "_").replace("/", "-")
        fname = f"{city_sanitized}_{_timestamp()}.json"
```

```
with open(day_dir / fname, "w", encoding="utf-8") as f:
    json.dump(data, f, ensure_ascii=False, indent=2)

results.append(data)
    time.sleep(pause_secs)

return results
```

Keep pause_secs to respect rate limits on free tiers.

8) Transform

etl/transform.py

```
from __future__ import annotations
from typing import List, Dict, Any
import pandas as pd
from datetime import datetime, timezone
# Utility to safely dig into nested dicts
def get(d: Dict[str, Any], path: str, default=None):
    cur = d
    for part in path.split("."):
        if not isinstance(cur, dict) or part not in cur:
            return default
        cur = cur[part]
    return cur
def normalize_current_weather(payloads: List[Dict[str, Any]]) -> pd.DataFrame:
    """Normalize OpenWeather current weather JSON payloads to a tidy DataFrame.
    Assumes units=metric unless noted otherwise.
    .....
    rows = []
    now_utc = datetime.now(timezone.utc).replace(microsecond=0)
    for p in payloads:
        rows.append({
            "city_id": get(p, "id"),
            "city_name": get(p, "name"),
            "country_code": get(p, "sys.country"),
            "lat": get(p, "coord.lat"),
            "lon": get(p, "coord.lon"),
```

```
# dt is Unix UTC timestamp of the observation
            "observation_time_utc": datetime.utcfromtimestamp(get(p, "dt",
0)).replace(microsecond=0),
            "temp_c": get(p, "main.temp"),
            "feels_like_c": get(p, "main.feels_like"),
            "temp_min_c": get(p, "main.temp_min"),
            "temp_max_c": get(p, "main.temp_max"),
            "pressure_hpa": get(p, "main.pressure"),
            "humidity_pct": get(p, "main.humidity"),
            "wind_speed_ms": get(p, "wind.speed"),
            "wind_deg": get(p, "wind.deg"),
            "clouds_pct": get(p, "clouds.all"),
            "weather_main": (get(p, "weather") or [{}])[0].get("main"),
            "weather_desc": (get(p, "weather") or [{}])[0].get("description"),
            "rain_1h_mm": get(p, "rain.1h", 0.0),
            "snow_1h_mm": get(p, "snow.1h", 0.0),
            "source": "openweather_current",
            "ingested_at_utc": now_utc,
        })
   df = pd.DataFrame(rows)
    # Enforce dtypes where reasonable
    dtype_map = {
        "city id": "Int64",
        "city_name": "string",
        "country_code": "string",
        "lat": "float64",
        "lon": "float64",
        "temp c": "float64",
        "feels_like_c": "float64",
        "temp_min_c": "float64",
        "temp_max_c": "float64",
        "pressure hpa": "Int64",
        "humidity_pct": "Int64",
        "wind speed ms": "float64",
        "wind deg": "Int64",
        "clouds_pct": "Int64",
        "weather_main": "string",
        "weather_desc": "string",
        "rain_1h_mm": "float64",
        "snow_1h_mm": "float64",
        "source": "string",
   }
    for col, dt in dtype map.items():
        if col in df.columns:
            df[col] = df[col].astype(dt)
```

```
# Datetime columns
for col in ["observation_time_utc", "ingested_at_utc"]:
    if col in df.columns:
        df[col] = pd.to_datetime(df[col], utc=True)

return df
```

9) Load

etl/load.py

```
from __future__ import annotations
from typing import Optional
from pathlib import Path
import pandas as pd
from sqlalchemy import create_engine, text
PROCESSED_DIR = Path("data/processed")
PROCESSED_DIR.mkdir(parents=True, exist_ok=True)
def save_as_csv(df: pd.DataFrame, basename: str) -> str:
   csv_path = PROCESSED_DIR / f"{basename}.csv"
   df.to_csv(csv_path, index=False)
   return str(csv_path)
def save_as_parquet(df: pd.DataFrame, basename: str) -> str:
   pq_path = PROCESSED_DIR / f"{basename}.parquet"
   df.to_parquet(pq_path, index=False)
    return str(pq_path)
def make_mysql_engine(user: str, password: str, host: str, port: int, database:
str):
   url = f"mysql+pymysql://{user}:{password}@{host}:{port}/{database}?
charset=utf8mb4"
   return create_engine(url, pool_pre_ping=True)
def ensure_table(engine, table_name: str):
    """Run a minimal DDL to ensure the target table exists.
    This is a safety net in case you didn't run sql/create_tables.sql.
```

```
.....
    ddl = f"""
    CREATE TABLE IF NOT EXISTS {table name} (
      id BIGINT PRIMARY KEY AUTO_INCREMENT,
      city_id BIGINT NULL,
      city_name VARCHAR(128) NOT NULL,
      country_code VARCHAR(8) NULL,
      lat DECIMAL(9,6) NULL,
      lon DECIMAL(9,6) NULL,
      observation time utc DATETIME NOT NULL,
      temp_c DECIMAL(6,2) NULL,
      feels like c DECIMAL(6,2) NULL,
      temp_min_c DECIMAL(6,2) NULL,
      temp_max_c DECIMAL(6,2) NULL,
      pressure_hpa INT NULL,
      humidity_pct INT NULL,
      wind_speed_ms DECIMAL(6,2) NULL,
      wind_deg INT NULL,
      clouds pct INT NULL,
      weather_main VARCHAR(64) NULL,
      weather_desc VARCHAR(256) NULL,
      rain_1h_mm DECIMAL(6,2) NULL,
      snow_1h_mm DECIMAL(6,2) NULL,
      source VARCHAR(64) NOT NULL,
      ingested_at_utc DATETIME NOT NULL,
      UNIQUE KEY uq_city_time (city_id, observation_time_utc)
    ) ENGINE=InnoDB;
    with engine.begin() as conn:
        conn.execute(text(ddl))
def append_to_mysql(df: pd.DataFrame, engine, table_name: str):
    # Pandas will map dtypes reasonably; for production, define explicit
SQLAlchemy types
    df.to sql(table name, con=engine, if exists='append', index=False,
method='multi', chunksize=1000)
```

10) Orchestrate the ETL (Manual Run)

etl/pipeline.py

```
from __future__ import annotations
import argparse
```

```
from pathlib import Path
import yaml
from datetime import datetime, timezone
from extract import extract_current_weather
from transform import normalize_current_weather
from load import save_as_csv, save_as_parquet, make_mysql_engine, ensure_table,
append_to_mysql
CONFIG_PATH = Path(__file__).resolve().parents[1] / 'config' / 'config.yaml'
def run_pipeline():
   # 1) Load config
   with open(CONFIG_PATH, 'r', encoding='utf-8') as f:
        cfg = yaml.safe_load(f)
   api_base = cfg['openweather']['api_base']
    api_key = cfg['openweather']['api_key']
   units = cfg['openweather'].get('units', 'metric')
   cities = cfg['run']['cities']
   save_parquet_flag = bool(cfg['run'].get('save_parquet', True))
   mysql_cfg = cfg.get('mysql', {})
   use_mysql = bool(mysql_cfg.get('enabled', False))
   # 2) Extract
   payloads = extract_current_weather(api_base, api_key, cities, units)
   # 3) Transform
   df = normalize_current_weather(payloads)
   # 4) Load → CSV/Parquet (always), and optionally MySQL
   ts = datetime.now(timezone.utc).strftime('%Y%m%dT%H%M%SZ')
   base = f"weather_current_{ts}"
   csv_path = save_as_csv(df, base)
   print(f"Saved CSV → {csv_path}")
   if save_parquet_flag:
        pq_path = save_as_parquet(df, base)
        print(f"Saved Parquet → {pq_path}")
   if use_mysql:
        engine = make_mysql_engine(
            mysql_cfg['user'], mysql_cfg['password'], mysql_cfg['host'],
```

11) Run It Manually

- 1) Put your API key and settings in config/config.yaml.
- 2) (Optional) Ensure the MySQL table exists (Section 6), or rely on the ensure_table() fallback.
- 3) Run the pipeline:

```
# from project root
python -m etl.pipeline
```

You should see:

- Raw JSON files in data/raw/YYYYMMDD/...
- A processed CSV (and Parquet) in data/processed/.
- If MySQL is enabled, new rows in weather weather observations .

12) Validate the Load

MySQL sample queries

```
SELECT COUNT(*) FROM weather_observations;

SELECT city_name, observation_time_utc, temp_c, humidity_pct
FROM weather_observations
ORDER BY observation_time_utc DESC
LIMIT 10;
```

13) CSV-Only Fallback (No MySQL)

If you don't have MySQL available yet, set in config.yaml:

```
mysql:
enabled: false
```

Your ETL will still generate data/processed/*.csv (and Parquet if enabled). You can load the CSV into Excel, pandas, or a BI tool.

14) Common Pitfalls & Fixes

- 401 Unauthorized: API key missing or wrong; confirm it's enabled for the Current Weather endpoint.
- **429 Too Many Requests**: Add a bigger pause_secs in extract_current_weather(); reduce number of cities.
- City not found: Ensure format City, CountryCode (e.g., Kathmandu, NP).
- Unicode errors on Windows: Always open files with encoding="utf-8" (already in code).
- **Duplicate rows**: The unique key (city_id, observation_time_utc) prevents exact dupes; if you still get errors, deduplicate in pandas before load.

15) Optional Enhancements (try later)

- Logging: Add logging to write to logs/pipeline.log instead of print.
- **Config via** . env : Keep secrets in environment variables; read them in code.
- Validation: Use pydantic models to validate payloads.
- **Tests**: Add a tests/ folder and write unit tests for transform logic.
- **PySpark version**: Swap pandas with PySpark DataFrame API once your dataset grows.
- Scheduling: Later add Airflow or cron for automation (outside this guide).

16) Learning & Reference Links

- OpenWeather product: *Current Weather Data* (check your account dashboard for exact URL and parameters)
- pandas docs: DataFrame, to_csv , to_parquet , to_sql
- SQLAlchemy docs: create engine strings for MySQL (dialect | mysql+pymysql)
- MySQL docs: data types, indexes, and UNIQUE KEY

(These are generic docs you can easily search; this quide stays tool-agnostic and copy-pasteable.)

17) What To Submit (if this is for an internship task)

- A link to your GitHub repo with:
- The project scaffold and code above
- A **README** explaining: what it does, how to run it, and sample outputs
- A screenshot of your MySQL table with a few rows (or the generated CSV)
- Short notes on: what went well, what you'd improve (e.g., retries, logging, tests)

You now have a clean, non-automated ETL you can run by hand.

When ready, we can iterate to a PySpark version and/or add Airflow for scheduling.