

Medallion Pipeline – Bronze Layer (Elhub Energy Data)

Scope of this notebook (BRONZE only):

- Retrieve hourly **production** and **consumption** data from the Elhub Energy Data API
 - Production: all price areas, years **2022–2024**
(PRODUCTION_PER_GROUP_MBA_HOUR)
 - Consumption: all price areas, years **2021–2024**
(CONSUMPTION_PER_GROUP_MBA_HOUR)
- Normalize the raw JSON into flat tabular form with minimal processing
- Store raw-ish data to:
 - Local bronze folder (CSV/Parquet)
 - Cassandra (bronze tables)
 - MongoDB (bronze collections)
- No business rules, no joins, no aggregations → that's for **silver** and **gold**.

```
In [ ]: # Imports & Global Config, for the bronze level of this compulsory assignment
import os
import time
import requests
from datetime import datetime, date, timedelta, timezone
import pandas as pd
import matplotlib.pyplot as plt

# Spark and DB switches
SPARK_AVAILABLE = False
try:
    from pyspark.sql import SparkSession
    from pyspark.sql.types import (
        StructType, StructField, StringType,
        TimestampType, DoubleType, IntegerType
    )
    SPARK_AVAILABLE = True
except Exception as e:
    print("Spark unavailable:", e)

print("Pandas version:", pd.__version__)
print("Spark available:", SPARK_AVAILABLE)

# Elhub API basic config
ELHUB_BASE = "https://api.elhub.no/energy-data/v0"
ENTITY = "price-areas"

DATASETS = {
    "production": "PRODUCTION_PER_GROUP_MBA_HOUR",
    "consumption": "CONSUMPTION_PER_GROUP_MBA_HOUR",
}
```

```

ELHUB_AUTH_TOKEN = None

# Time ranges (bronze)
PROD_START = date(2022, 1, 1)
PROD_END   = date(2024, 12, 31)

CONS_START = date(2021, 1, 1)
CONS_END   = date(2024, 12, 31)

# Setting the local storage paths for the bronze level
BRONZE_ROOT = "../Data_Assignment_4/bronze"
BRONZE_PROD_DIR = os.path.join(BRONZE_ROOT, "production")
BRONZE_CONS_DIR = os.path.join(BRONZE_ROOT, "consumption")

os.makedirs(BRONZE_PROD_DIR, exist_ok=True)
os.makedirs(BRONZE_CONS_DIR, exist_ok=True)

# Cassandra & Mongo config (bronze)
CASSANDRA_HOSTS = "127.0.0.1"
CASSANDRA_KEYSPACE = "energy"
CASSANDRA_TABLE_BRONZE_PROD = "bronze_production_hourly"
CASSANDRA_TABLE_BRONZE_CONS = "bronze_consumption_hourly"

MONGO_URI = "mongodb://localhost:27017"
MONGO_DATABASE = "ind320"
MONGO_COLL_BRONZE_PROD = "bronze_production_hourly"
MONGO_COLL_BRONZE_CONS = "bronze_consumption_hourly"

```

Spark unavailable: No module named 'pyspark.errors'
Pandas version: 2.3.3
Spark available: False

```

In [ ]: # We set the API headers here
def elhub_headers():
    headers = {
        "Accept": "application/vnd.api+json, application/json",
    }
    if ELHUB_AUTH_TOKEN:
        headers["Authorization"] = ELHUB_AUTH_TOKEN
    return headers

# We return the ISO 8601 UTC format for datetime objects
def iso_utc(d: datetime) -> str:
    if d.tzinfo is None:
        d = d.replace(tzinfo=timezone.utc)
    else:
        d = d.astimezone(timezone.utc)
    return d.isoformat().replace("+00:00", "Z")

# Then we ensure that the response is JSON
def ensure_json(r: requests.Response):
    ct = (r.headers.get("Content-Type") or "").lower()
    if "json" not in ct:
        raise ValueError(f"Non-JSON response (status {r.status_code}) from {r.url}")
    return r.json()

```

```

# Here we fetch data from Elhub for a given dataset and time interval
def fetch_interval_json(dataset_key: str, start_dt: datetime, end_dt: datetime):
    dataset = DATASETS[dataset_key]
    params = {
        "dataset": dataset,
        "startTime": iso_utc(start_dt),
        "endTime": iso_utc(end_dt),
        "page[size]": 10000,
    }
    url = f"{ELHUB_BASE}/{ENTITY}"
    r = requests.get(url, headers=elhub_headers(), params=params, timeout=120)
    if r.status_code != 200:
        raise RuntimeError(
            f"Elhub API error {r.status_code} for {dataset_key}: {r.text[:500]}"
        )
    payload = ensure_json(r)
    return payload

```

```

In [ ]: # Now we normalize the Elhub JSON payload into a pandas DataFrame
def normalize_elhub_items(payload: dict, kind: str) -> pd.DataFrame:
    data = payload.get("data") or []
    rows = []

    for item in data:
        attrs = (item or {}).get("attributes") or {}
        prod_series = attrs.get("productionPerGroupMbaHour") or []
        cons_series = attrs.get("consumptionPerGroupMbaHour") or []

        # Generic handler for series lists
        def handle_series(series, is_production: bool):
            for e in series:
                row = {}
                row["priceArea"] = e.get("priceArea") or attrs.get("name")

                if is_production:
                    row["group"] = e.get("productionGroup")
                    row["quantityKwh"] = e.get("quantityKwh")
                    row["countMeteringPoints"] = None
                else:
                    row["group"] = e.get("consumptionGroup")
                    row["quantityKwh"] = e.get("quantityKwh")
                    row["countMeteringPoints"] = e.get("countMeteringPoints")

                row["startTime"] = e.get("startTime")
                row["endTime"] = e.get("endTime")
                rows.append(row)

        if prod_series and kind == "production":
            handle_series(prod_series, is_production=True)
        if cons_series and kind == "consumption":
            handle_series(cons_series, is_production=False)

    # Fallback: sometimes a single timeseries dict instead of list
    ts = attrs.get("timeseries") or {}
    if ts and not (prod_series or cons_series):

```

```

        row = {}
        row["priceArea"] = ts.get("priceArea") or attrs.get("name")
        if kind == "production":
            row["group"] = ts.get("productionGroup")
            row["quantityKwh"] = ts.get("quantityKwh")
            row["countMeteringPoints"] = None
        else:
            row["group"] = ts.get("consumptionGroup")
            row["quantityKwh"] = ts.get("quantityKwh")
            row["countMeteringPoints"] = ts.get("countMeteringPoints")

        row["startTime"] = ts.get("startTime")
        row["endTime"] = ts.get("endTime")
        rows.append(row)

    if not rows:
        return pd.DataFrame(
            columns=["priceArea", "group", "startTime", "endTime",
                    "quantityKwh", "countMeteringPoints"]
        )

    df = pd.DataFrame(rows)

    # Light typing (still bronze-friendly)
    df["startTime"] = pd.to_datetime(df["startTime"], errors="coerce", utc=True)
    df["endTime"] = pd.to_datetime(df["endTime"], errors="coerce", utc=True)
    df["quantityKwh"] = pd.to_numeric(df["quantityKwh"], errors="coerce")
    if "countMeteringPoints" in df.columns:
        df["countMeteringPoints"] = pd.to_numeric(df["countMeteringPoints"], errors="coerce")

    return df

```

```

In [ ]: # Helper to iterate months in a date range
def iter_months(start_d: date, end_d: date):
    cur = date(start_d.year, start_d.month, 1)
    while cur <= end_d:
        # next month
        if cur.month == 12:
            nxt = date(cur.year + 1, 1, 1)
        else:
            nxt = date(cur.year, cur.month + 1, 1)
        yield cur, min(nxt - timedelta(days=1), end_d)
        cur = nxt

# We fetch & normalize data for given kind ('production' or 'consumption')
def fetch_dataset_range(kind: str, start_d: date, end_d: date, pause: float = 1.0):
    assert kind in ("production", "consumption")
    frames = []
    failures = []

    for month_start, month_end in iter_months(start_d, end_d):
        start_dt = datetime.combine(month_start, datetime.min.time()).replace(tzinfo=timezone.utc)
        end_dt = datetime.combine(month_end + timedelta(days=1), datetime.min.time()).replace(tzinfo=timezone.utc)

        label = f"{month_start:%Y-%m}"

```

```

print(f"Fetching {kind} for {label} ({start_dt} → {end_dt}) ...", end=" ")

try:
    payload = fetch_interval_json(kind, start_dt, end_dt)
    df_month = normalize_elhub_items(payload, kind=kind)
    print(f"{len(df_month):,} rows")
    if not df_month.empty:
        frames.append(df_month)
except Exception as e:
    print("FAILED:", e)
    failures.append((label, str(e)))

time.sleep(pause)

if frames:
    df_all = pd.concat(frames, ignore_index=True)
else:
    df_all = pd.DataFrame(
        columns=["priceArea", "group", "startTime", "endTime",
                "quantityKwh", "countMeteringPoints"]
    )

return df_all, failures

```

```

In [ ]: # Spark-related functions for the bronze level
def build_spark(app_name="Elhub Bronze Ingestion"):
    if not SPARK_AVAILABLE:
        raise RuntimeError("Spark is not available in this environment.")
    spark = (
        SparkSession.builder
        .appName(app_name)
        .config("spark.cassandra.connection.host", CASSANDRA_HOSTS)
        .config("spark.mongodb.write.connection.uri", MONGO_URI)
        .getOrCreate()
    )
    return spark

def to_spark(df: pd.DataFrame):
    if not SPARK_AVAILABLE:
        raise RuntimeError("Spark is not available.")
    schema = StructType([
        StructField("priceArea", StringType(), True),
        StructField("group", StringType(), True),
        StructField("startTime", TimestampType(), True),
        StructField("endTime", TimestampType(), True),
        StructField("quantityKwh", DoubleType(), True),
        StructField("countMeteringPoints", IntegerType(), True),
    ])
    spark = build_spark()
    return spark.createDataFrame(df, schema=schema)

def write_bronze_to_cassandra(sdf, table_name: str):
    (
        sdf.write

```

```

        .format("org.apache.spark.sql.cassandra")
        .mode("append")
        .options(keyspace=CASSANDRA_KEYSPACE, table=table_name)
        .save()
    )

def write_bronze_to_mongo(sdf, collection_name: str):
    (
        sdf.write
        .format("mongodb")
        .mode("append")
        .option("database", MONGO_DATABASE)
        .option("collection", collection_name)
        .save()
    )

```

```

In [ ]: # Now we perform the bronze ingestion: PRODUCTION 2022-2024
prod_df, prod_fail = fetch_dataset_range("production", PROD_START, PROD_END)

print("\nProduction fetch done.")
print("Rows:", len(prod_df))
print("Failures:", prod_fail[:5])

# And we save the production data locally, partitioned by year
if not prod_df.empty:
    prod_df = prod_df.copy()
    prod_df["year"] = prod_df["startTime"].dt.year
    print("Years present in production:", sorted(prod_df["year"].dropna().unique()))

    for year, df_y in prod_df.groupby("year"):
        out_path = os.path.join(BRONZE_PROD_DIR, f"production_{int(year)}.csv")
        df_y.drop(columns=["year"]).to_csv(out_path, index=False)
        print(f"Saved {len(df_y):,} rows to {out_path}")
else:
    print("⚠ No production rows fetched - nothing saved.")

```

[illegible]

```

Fetching production for 2024-05 (2024-05-01 00:00:00+00:00 → 2024-06-01 00:00:00+00:00) ... 18,851 rows
Fetching production for 2024-06 (2024-06-01 00:00:00+00:00 → 2024-07-01 00:00:00+00:00) ... 18,851 rows
Fetching production for 2024-07 (2024-07-01 00:00:00+00:00 → 2024-08-01 00:00:00+00:00) ... 18,851 rows
Fetching production for 2024-08 (2024-08-01 00:00:00+00:00 → 2024-09-01 00:00:00+00:00) ... 18,851 rows
Fetching production for 2024-09 (2024-09-01 00:00:00+00:00 → 2024-10-01 00:00:00+00:00) ... 18,851 rows
Fetching production for 2024-10 (2024-10-01 00:00:00+00:00 → 2024-11-01 00:00:00+00:00) ... 18,851 rows
Fetching production for 2024-11 (2024-11-01 00:00:00+00:00 → 2024-12-01 00:00:00+00:00) ... 18,851 rows
Fetching production for 2024-12 (2024-12-01 00:00:00+00:00 → 2025-01-01 00:00:00+00:00) ... 18,851 rows

```

Production fetch done.

Rows: 678636

Failures: []

Years present in production: [np.int32(2025)]

Saved 678,636 rows to ../Data_Assignment_4/bronze\production\production_2025.csv

```

In [ ]: # We do some quick checks on the production DataFrame
print("prod_df dtypes:\n", prod_df.dtypes)
print("\nstartTime sample:")
print(prod_df["startTime"].head())

if not prod_df.empty:
    print("\nUnique years in startTime:")
    print(sorted(prod_df["startTime"].dt.year.dropna().unique()))

```

```

prod_df dtypes:
priceArea          object
group              object
quantityKwh        float64
countMeteringPoints float64
startTime          datetime64[ns, UTC]
endTime            datetime64[ns, UTC]
year               int32
dtype: object

```

```

startTime sample:
0   2025-10-22 19:00:00+00:00
1   2025-10-22 20:00:00+00:00
2   2025-10-22 21:00:00+00:00
3   2025-10-22 22:00:00+00:00
4   2025-10-22 23:00:00+00:00
Name: startTime, dtype: datetime64[ns, UTC]

```

```

Unique years in startTime:
[np.int32(2025)]

```

```

In [ ]: # and we do the same for the CONSUMPTION data: 2021-2024
cons_df, cons_fail = fetch_dataset_range("consumption", CONS_START, CONS_END)

```



```
print("\nConsumption fetch done.")
print("Rows:", len(cons_df))
print("Failures:", cons_fail[:5])

# And we save the consumption data locally, partitioned by year
if not cons_df.empty:
    cons_df = cons_df.copy()
    cons_df["year"] = cons_df["startTime"].dt.year
    print("Years present in consumption:", sorted(cons_df["year"].dropna().unique()))

    for year, df_y in cons_df.groupby("year"):
        out_path = os.path.join(BRONZE_CONS_DIR, f"consumption_{int(year)}.csv")
        df_y.drop(columns=["year"]).to_csv(out_path, index=False)
        print(f"Saved {len(df_y):,} rows to {out_path}")
else:
    print("⚠ No consumption rows fetched - nothing saved.")
```

[illegible]

Fetching consumption for 2023-05 (2023-05-01 00:00:00+00:00 → 2023-06-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2023-06 (2023-06-01 00:00:00+00:00 → 2023-07-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2023-07 (2023-07-01 00:00:00+00:00 → 2023-08-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2023-08 (2023-08-01 00:00:00+00:00 → 2023-09-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2023-09 (2023-09-01 00:00:00+00:00 → 2023-10-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2023-10 (2023-10-01 00:00:00+00:00 → 2023-11-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2023-11 (2023-11-01 00:00:00+00:00 → 2023-12-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2023-12 (2023-12-01 00:00:00+00:00 → 2024-01-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-01 (2024-01-01 00:00:00+00:00 → 2024-02-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-02 (2024-02-01 00:00:00+00:00 → 2024-03-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-03 (2024-03-01 00:00:00+00:00 → 2024-04-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-04 (2024-04-01 00:00:00+00:00 → 2024-05-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-05 (2024-05-01 00:00:00+00:00 → 2024-06-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-06 (2024-06-01 00:00:00+00:00 → 2024-07-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-07 (2024-07-01 00:00:00+00:00 → 2024-08-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-08 (2024-08-01 00:00:00+00:00 → 2024-09-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-09 (2024-09-01 00:00:00+00:00 → 2024-10-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-10 (2024-10-01 00:00:00+00:00 → 2024-11-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-11 (2024-11-01 00:00:00+00:00 → 2024-12-01 00:00:00+00:00) ... 18,100 rows
 Fetching consumption for 2024-12 (2024-12-01 00:00:00+00:00 → 2025-01-01 00:00:00+00:00) ... 18,100 rows

Consumption fetch done.

Rows: 868800

Failures: []

Years present in consumption: [np.int32(2025)]

Saved 868,800 rows to ../Data_Assignment_4/bronze\consumption\consumption_2025.csv

```

In [ ]: # We do some quick checks on the consumption DataFrame
        # Checking for consumption DataFrame
        if not prod_df.empty:
            print("Production time span:",
                  prod_df["startTime"].min(), "→", prod_df["startTime"].max())
            print(prod_df.head())

        if not cons_df.empty:
            print("Consumption time span:",
  
```

```
cons_df["startTime"].min(), ">", cons_df["startTime"].max())
print(cons_df.head())
```

Production time span: 2025-10-22 19:00:00+00:00 → 2025-11-21 22:00:00+00:00

	priceArea	group	quantityKwh	countMeteringPoints	\
0	NO1	hydro	2332513.0		NaN
1	NO1	hydro	2281520.8		NaN
2	NO1	hydro	2308167.5		NaN
3	NO1	hydro	2253987.5		NaN
4	NO1	hydro	2241552.0		NaN

		startTime		endTime	year
0	2025-10-22	19:00:00+00:00	2025-10-22	20:00:00+00:00	2025
1	2025-10-22	20:00:00+00:00	2025-10-22	21:00:00+00:00	2025
2	2025-10-22	21:00:00+00:00	2025-10-22	22:00:00+00:00	2025
3	2025-10-22	22:00:00+00:00	2025-10-22	23:00:00+00:00	2025
4	2025-10-22	23:00:00+00:00	2025-10-23	00:00:00+00:00	2025

Consumption time span: 2025-10-22 19:00:00+00:00 → 2025-11-21 22:00:00+00:00

	priceArea	group	quantityKwh	countMeteringPoints	\
0	NO1	cabin	67069.060		NaN
1	NO1	cabin	66767.110		NaN
2	NO1	cabin	65964.305		NaN
3	NO1	cabin	65159.824		NaN
4	NO1	cabin	64141.660		NaN

		startTime		endTime	year
0	2025-10-22	19:00:00+00:00	2025-10-22	20:00:00+00:00	2025
1	2025-10-22	20:00:00+00:00	2025-10-22	21:00:00+00:00	2025
2	2025-10-22	21:00:00+00:00	2025-10-22	22:00:00+00:00	2025
3	2025-10-22	22:00:00+00:00	2025-10-22	23:00:00+00:00	2025
4	2025-10-22	23:00:00+00:00	2025-10-23	00:00:00+00:00	2025