Pipeline E2E AWS per analizzare BTC e XMR

Introduzione

CryptoData Insights è un'azienda innovativa specializzata nell'analisi dei dati sulle criptovalute. Questo progetto implementa una pipeline End-to-End (E2E) su AWS per analizzare Bitcoin (BTC) e Monero (XMR). La pipeline automatizza il processo di acquisizione, pulizia e trasformazione dei dati, rendendoli pronti per l'esplorazione e la visualizzazione.

Architettura della Soluzione

La pipeline si articola in diverse fasi:

- 1. Caricamento dati grezzi su S3.
- 2. Pulizia e trasformazione dei dati da Raw a Silver.
- 3. Trasformazione dei dati da Silver a Golden.
- 4. Creazione dei Glue Jobs
- 5. Creazione gruppo di lavoro Redshift
- 6. Caricamento su Redshift da Golden.
- 7. Orchestrazione con Step Functions per coordinare tutte le fasi.

Di seguito sono descritti tutti i passaggi implementati.

Passaggi Implementati

Creazione dei Bucket S3

```
CloudShell

eu-north-1 +

* aws s3 mb s3://crypto-raw-bucket
make_bucket: crypto-raw-bucket

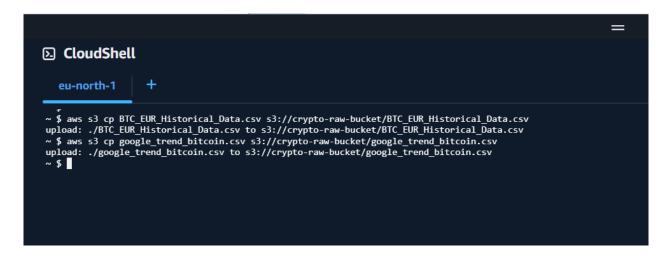
* aws s3 mb s3://crypto-silver
make_bucket: crypto-silver

* aws s3 mb s3://crypto-golden
make bucket: crypto-golden

* $ aws s3 mb s3://crypto-golden

* $ aws s3 mb s3://crypto-golden
```

Caricamento dei File CSV su S3



Creazione degli Script per Trasformazioni ETL

raw-silver-btc.py

```
import sys
from awsglue.utils import getResolvedOptions
from pyspark.context import SparkContext
from awsglue.context import GlueContext
from awsglue.dynamicframe import DynamicFrame
from pyspark.sql import functions as F
from pyspark.sql.functions import col, to date, regexp replace, round
from pyspark.sql.window import Window
# Parametri
args = getResolvedOptions(sys.argv, ['JOB_NAME'])
sc = SparkContext()
glueContext = GlueContext(sc)
spark = glueContext.spark_session
# --- Path configurati ---
RAW BUCKET = "s3://crypto-raw-bucket"
SILVER BUCKET = "s3://crypto-silver"
BTC RAW PATH = f"{RAW BUCKET}/BTC EUR Historical Data.csv"
GT BTC RAW PATH = f"{RAW BUCKET}/google trend bitcoin.csv"
BTC SILVER PATH = f"{SILVER BUCKET}/btc"
GT_BTC_SILVER_PATH = f"{SILVER_BUCKET}/gt_bitcoin"
# --- Funzioni ETL ---
```

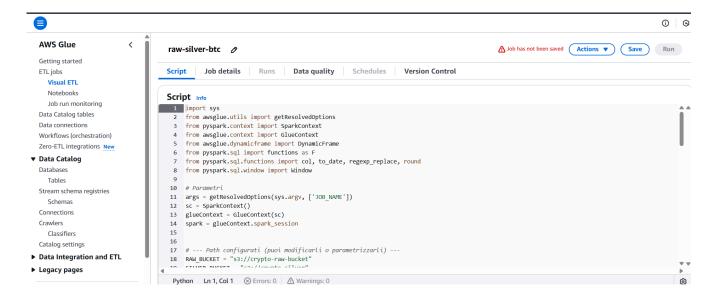
```
def load csv(path: str):
  dynf = glueContext.create dynamic frame.from options(
    connection type="s3",
    connection_options={"paths": [path]},
    format="csv",
    format_options={"withHeader": True}
  )
  return dynf.toDF()
def clean price df(df, date col="Date", price col="Price"):
  df = df.withColumn(price_col, regexp_replace(col(price_col), ",", "").cast("float"))
  df = df.withColumn(date col, to date(col(date col), "MM/dd/yyyy"))
  window = Window.orderBy(date col)
  df = df.withColumn(
    price col,
    F.when((col(price col).isNull()) | (col(price col) == -1),
        (F.lag(price col).over(window) + F.lead(price col).over(window)) / 2)
    .otherwise(col(price col))
  )
  df = df.withColumn(price_col, round(col(price_col), 3))
  return df
def clean trend df(df, date col="Settimana", value col="interesse bitcoin"):
  return df.withColumn(date col, col(date col).cast("date")) \
       .withColumn(value col, col(value col).cast("float"))
def save parquet(df, output path):
  dynf = DynamicFrame.fromDF(df, glueContext, "parquet out")
  glueContext.write dynamic frame.from options(
    frame=dvnf,
    connection type="s3",
    connection_options={"path": output_path},
    format="parquet"
  )
# --- Esecuzione ETL ---
btc_raw_df = load_csv(BTC_RAW_PATH)
gt_btc_raw_df = load_csv(GT_BTC_RAW_PATH)
btc_clean_df = clean_price_df(btc_raw_df)
gt btc clean df = clean trend df(gt btc raw df)
save parquet(btc clean df, BTC SILVER PATH)
save parquet(gt btc clean df, GT BTC SILVER PATH)
```

silver-gold-btc.py

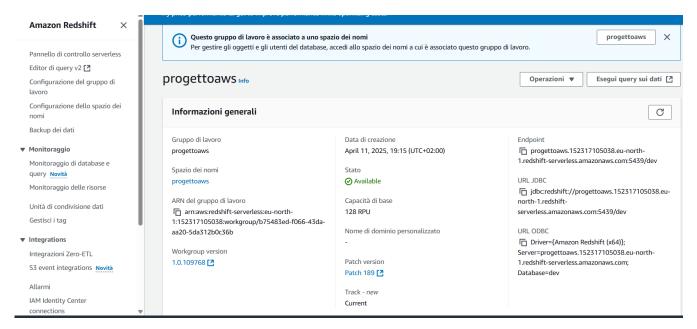
```
import sys
from awsglue.utils import getResolvedOptions
from pyspark.context import SparkContext
from awsglue.context import GlueContext
from awsglue.dynamicframe import DynamicFrame
from pyspark.sql.functions import col, avg
from pyspark.sql.window import Window
# Parametri
args = getResolvedOptions(sys.argv, ['JOB NAME'])
sc = SparkContext()
glueContext = GlueContext(sc)
spark = glueContext.spark session
# --- Percorsi Silver e Gold ---
SILVER BUCKET = "s3://crypto-silver"
GOLD BUCKET = "s3://crypto-golden"
BTC SILVER PATH = f"{SILVER BUCKET}/btc"
GT_BTC_SILVER_PATH = f"{SILVER_BUCKET}/gt_bitcoin"
BTC_GOLD_PATH = f"{GOLD_BUCKET}/btc_with_trend"
# --- Funzioni di supporto ---
def load parquet(path):
  dynf = glueContext.create dynamic frame.from options(
    connection type="s3",
    connection_options={"paths": [path]},
    format="parquet"
  )
  return dynf.toDF()
def save to gold(df, path):
  dynf = DynamicFrame.fromDF(df, glueContext, "gold_output")
  glueContext.write dynamic frame.from options(
    frame=dynf,
    connection type="s3",
    connection_options={"path": path},
    format="parquet"
  )
# --- ETL: Caricamento dati Silver ---
btc_df = load_parquet(BTC_SILVER_PATH)
gt df = load parquet(GT BTC SILVER PATH)
# --- Media mobile a 10 giorni sul prezzo BTC ---
window spec = Window.orderBy("Date").rowsBetween(-9, 0) # 10 giorni (incluso il corrente)
btc_df = btc_df.withColumn("Price", avg("Price").over(window_spec))
```

Gli script per Monero (raw-silver-xmr.py e silver-gold-xmr.py) sono analoghi.

Creazione dei Glue Jobs



Creazione del Gruppo di Lavoro su Redshift

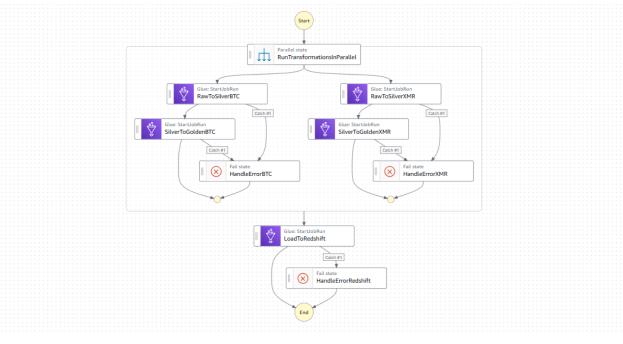


Script per il Caricamento dei Dati su Redshift

```
import psycopg2
import pandas as pd
from pyspark.sql import SparkSession
# Parametri di connessione
JDBC URL = "jdbc:redshift://progettoaws.152317105038.eu-north-1.redshift-
serverless.amazonaws.com:5439/dev"
USER = "admin"
PASSWORD = "JIMFDInbt792"
PORT = "5439"
DATABASE = "dev"
# Funzione per connettersi a Redshift e caricare i dati
def connect_to_redshift():
  conn = psycopg2.connect(
    dbname=DATABASE,
    user=USER,
    password=PASSWORD,
    host="progettoaws.152317105038.eu-north-1.redshift-serverless.amazonaws.com",
    port=PORT
  )
  return conn
# Caricamento dei dati da S3 (per Bitcoin e Monero)
def load data from s3 to redshift():
  # Inizializza una sessione Spark per caricare i dati da Parquet su S3
  spark = SparkSession.builder.appName("S3-to-Redshift").getOrCreate()
  # Carica i dati da S3 per Bitcoin e Monero (nel bucket crypto-golden)
  btc_df = spark.read.parquet("s3://crypto-golden/btc_with_ trend")
  xmr_df = spark.read.parquet("s3://crypto-golden/xmr_with_trend")
  # Converte i DataFrame Spark in Pandas DataFrame per inviarli a Redshift
  btc_pd = btc_df.toPandas()
  xmr pd = xmr df.toPandas()
  # Connessione a Redshift
  conn = connect to redshift()
  cursor = conn.cursor()
  # Creazione delle tabelle (se non esistono già)
  create_btc_table_query = """
  CREATE TABLE IF NOT EXISTS btc_with_trend (
    Date DATE,
    Price FLOAT,
    GoogleTrend FLOAT
```

```
);
  cursor.execute(create btc table query)
  create_xmr_table_query = """
  CREATE TABLE IF NOT EXISTS xmr_with_trend (
    Date DATE,
    Price FLOAT,
    GoogleTrend FLOAT
  );
  cursor.execute(create xmr table query)
  conn.commit()
  # Inserimento dei dati da Pandas DataFrame a Redshift per Bitcoin
  for index, row in btc pd.iterrows():
    insert btc query = """
    INSERT INTO btc_with_trend (Date, Price, GoogleTrend)
    VALUES (%s, %s, %s)
    cursor.execute(insert_btc_query, (row['Date'], row['Price'], row['GoogleTrend']))
  # Inserimento dei dati da Pandas DataFrame a Redshift per Monero
  for index, row in xmr pd.iterrows():
    insert_xmr_query = """
    INSERT INTO xmr with trend (Date, Price, GoogleTrend)
    VALUES (%s, %s, %s)
    cursor.execute(insert_xmr_query, (row['Date'], row['Price'], row['GoogleTrend']))
  conn.commit()
  # Chiude la connessione
  cursor.close()
  conn.close()
# Esegui il caricamento
load_data_from_s3_to_redshift()
```

Definizione delle Step Functions



```
{
  "Comment": "Pipeline E2E per BTC e XMR",
  "StartAt": "RunTransformationsInParallel",
  "States": {
   "RunTransformationsInParallel": {
    "Type": "Parallel",
    "Branches": [
      "StartAt": "RawToSilverBTC",
      "States": {
       "RawToSilverBTC": {
        "Type": "Task",
        "Resource": "arn:aws:states:::glue:startJobRun.sync",
        "Parameters": {
          "JobName": "raw-silver-btc"
        "Next": "SilverToGoldenBTC",
        "Catch": [
           "ErrorEquals": [
            "States.ALL"
           "Next": "HandleErrorBTC"
        ]
        "SilverToGoldenBTC": {
        "Type": "Task",
        "Resource": "arn:aws:states:::glue:startJobRun.sync",
```

```
"Parameters": {
    "JobName": "silver-gold-btc"
   },
   "End": true,
   "Catch": [
     "ErrorEquals": [
      "States.ALL"
     "Next": "HandleErrorBTC"
    }
   ]
  },
  "HandleErrorBTC": {
   "Type": "Fail",
   "Error": "BTCPipelineFailed",
   "Cause": "Errore nella pipeline BTC"
}
},
 "StartAt": "RawToSilverXMR",
 "States": {
  "RawToSilverXMR": {
   "Type": "Task",
   "Resource": "arn:aws:states:::glue:startJobRun.sync",
   "Parameters": {
    "JobName": "raw-silver-xmr"
   "Next": "SilverToGoldenXMR",
   "Catch": [
     "ErrorEquals": [
      "States.ALL"
     "Next": "HandleErrorXMR"
    }
   ]
  },
  "SilverToGoldenXMR": {
   "Type": "Task",
   "Resource": "arn:aws:states:::glue:startJobRun.sync",
   "Parameters": {
    "JobName": "silver-gold-xmr"
   },
   "End": true,
   "Catch": [
    {
```

```
"ErrorEquals": [
          "States.ALL"
         "Next": "HandleErrorXMR"
      1
     },
      "HandleErrorXMR": {
       "Type": "Fail",
       "Error": "XMRPipelineFailed",
       "Cause": "Errore nella pipeline XMR"
    }
   }
  "Next": "LoadToRedshift"
 "LoadToRedshift": {
  "Type": "Task",
  "Resource": "arn:aws:states:::glue:startJobRun.sync",
  "Parameters": {
   "JobName": "load_redshift"
  },
  "End": true,
  "Catch": [
    "ErrorEquals": [
     "States.ALL"
    "Next": "HandleErrorRedshift"
   }
 },
 "HandleErrorRedshift": {
  "Type": "Fail",
  "Error": "RedshiftLoadFailed",
  "Cause": "Errore durante il caricamento in Redshift"
 }
}
```

Risultati Attesi

Al termine della pipeline:

- I dati grezzi vengono puliti e trasformati.
- I dati finali (prezzo e trend unificati) vengono caricati su Redshift.
- Le tabelle **btc_with_trend** e **xmr_with_trend** sono pronte per l'analisi SQL o la visualizzazione con QuickSight.

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