

# Big Data 3

PROJEKAT 3

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# Dataset - US Accidents (2016 - 2023)

- 1 Skup podataka o saobraćajnim nezgodama u 49 država u SAD.
- 2 Podaci su prikupljeni od februara 2016. do marta 2023. godine.
- 3 Dataset sadrži oko 7 miliona zapisa i veličine je 3.06 GB. Postoji i umanjena verzija sa oko 500 000 zapisa i veličine 188 MB. Dataset ima 46 kolona.

# Analiza dataseta

- 1 Dataset sadrži veliki broj null vrednosti u različitim kolonama.
- 2 Takođe pojedine vrednosti kolona koje se odnose na vreme nisu standardizovane.
- 3 Pojedine kolone su preimenovane zbog lakšeg rada

# preprocessor.py

```
import pandas as pd

df = pd.read_csv('us_accidents.csv')

df.dropna()

df['Start_Time'] = pd.to_datetime(df['Start_Time'], errors='coerce')
df['End_Time'] = pd.to_datetime(df['End_Time'], errors='coerce')

rename_dict = {
    'Distance(mi)': 'Distance_mi',
    'Temperature(F)': 'Temperature_F',
    'Wind_Chill(F)': 'Wind_Chill_F',
    'Humidity(%)': 'Humidity_percent',
    'Pressure(in)': 'Pressure_in',
    'Visibility(mi)': 'Visibility_mi',
    'Wind_Speed(mph)': 'Wind_Speed_mph',
    'Precipitation(in)': 'Precipitation_in'
}

df = df.rename(columns=rename_dict)

df.to_csv('us_accidents_cleaned.csv', index=False)
```

# producer.py

```
from confluent_kafka import Producer
from uuid import uuid4
import time

config = {
    'bootstrap.servers': 'localhost:9092'
}

producer = Producer(config)

file = open('us_accidents_cleaned.csv', 'r', encoding='utf-8')

line = file.readline()
for line in file:
    producer.produce(
        topic = 'us_accidents_topic',
        key = str(uuid4()),
        value = line.strip()
    )

    print(line)
    producer.flush()
    time.sleep(3)

file.close()
```

# trainer.py

```
from pyspark.sql import SparkSession
from pyspark.sql.functions import col, when
from pyspark.ml import Pipeline
from pyspark.ml.feature import StringIndexer, VectorAssembler
from pyspark.ml.classification import GBTClassifier
from pyspark.ml.evaluation import BinaryClassificationEvaluator

spark = SparkSession.builder \
    .appName("AccidentDataTrainer") \
    .getOrCreate()

df = spark.read.csv(
    "us_accidents_cleaned.csv",
    header=True,
    inferSchema=True
)
```

```
df = df.withColumn(
    "label",
    when(col("Severity") >= 3, 1).otherwise(0)
)

fractions = {0: 0.06, 1: 1.0}
df = df.sampleBy("label", fractions, seed=36)

train_df, test_df = df.randomSplit([0.85, 0.15], seed=5043)
```

```
categorical_cols = [
    "Weather_Condition",
    "Wind_Direction",
    "Sunrise_Sunset"
]

indexers = [
    StringIndexer(
        inputCol=c,
        outputCol=c + "Indexed",
        handleInvalid="keep"
    )
    for c in categorical_cols
]
```

# trainer.py

```
feature_cols = [
    "Visibility_mi", "Wind_Speed_mph",
    "Temperature_F", "Humidity_percent", "Pressure_in",
    "Distance_mi", "Precipitation_in",
    "Start_Lat", "Start_Lng",

    "Junction", "Traffic_Signal", "Crossing",

    "Weather_ConditionIndexed",
    "Wind_DirectionIndexed",
    "Sunrise_SunsetIndexed"
]

assembler = VectorAssembler(
    inputCols=feature_cols,
    outputCol="features",
    handleInvalid="skip"
)

gbt = GBTClassifier(
    featuresCol="features",
    labelCol="label",
    maxIter=30,
    maxDepth=5,
    maxBins=150,
    seed=36
)

pipeline = Pipeline(
    stages=indexers + [assembler, gbt]
)
```

```
model = pipeline.fit(train_df)

predictions = model.transform(test_df)

predictions.select(
    "label", "prediction", "probability"
).show(5, truncate=False)

evaluator = BinaryClassificationEvaluator(
    labelCol="label",
    rawPredictionCol="rawPrediction",
    metricName="areaUnderROC"
)

auc = evaluator.evaluate(predictions)
print(f"Procena modela: {auc}")

model.write().overwrite().save("models/accident_gbt_pipeline")
```

Procena modela:  
0.8266998013514448



# consumer.py

```
from pyspark.sql import SparkSession
from pyspark.sql.types import StructType, StructField, StringType, IntegerType, DoubleType, BooleanType, TimestampType
from pyspark.sql.functions import from_csv, avg, window, col, count, collect_list, udf
from pyspark.ml import PipelineModel
from collections import Counter
import sys

window_type = sys.argv[1]
window_duration = sys.argv[2]
slide_duration = sys.argv[3]

spark = SparkSession.builder.appName("Big-Data-3") \
    .master("local[*]") \
    .getOrCreate()

model = PipelineModel.load("models/accident_gbt_pipeline")
```

```
string_schema = accidents_schema.simpleString()

df = (
    spark.readStream.format("kafka") \
        .option("kafka.bootstrap.servers", "localhost:9092") \
        .option("subscribe", "us_accidents_topic") \
        .option("startingOffsets", "earliest") \
        .load() \
        .selectExpr("CAST(value AS STRING) as csv_line") \
)

parsed_df = df.select(from_csv(df.csv_line, string_schema).alias("data")).select("data.*")

if (window_type.lower() == "tumbling"):
    windowed_col = window(col("Start_Time"), window_duration)
else:
    windowed_col = window(col("Start_Time"), window_duration, slide_duration)
```

```
accidents_schema = StructType([
    StructField("ID", StringType(), True),
    StructField("Source", StringType(), True),
    StructField("Severity", IntegerType(), True),
    StructField("Start_Time", TimestampType(), True),
    StructField("End_Time", TimestampType(), True),
    StructField("Start_Lat", DoubleType(), True),
    StructField("Start_Lng", DoubleType(), True),
    StructField("End_Lat", DoubleType(), True),
    StructField("End_Lng", DoubleType(), True),
    StructField("Distance_mi", DoubleType(), True),
    StructField("Description", StringType(), True),
    StructField("Street", StringType(), True),
    StructField("City", StringType(), True),
    StructField("County", StringType(), True),
    StructField("State", StringType(), True),
    StructField("Zipcode", StringType(), True),
    StructField("Country", StringType(), True),
    StructField("Timezone", StringType(), True),
    StructField("Airport_Code", StringType(), True),
    StructField("Weather_Timestamp", TimestampType(), True),
    StructField("Temperature_F", DoubleType(), True),
    StructField("Wind_Chill_F", DoubleType(), True),
    StructField("Humidity_percent", DoubleType(), True),
    StructField("Pressure_in", DoubleType(), True),
    StructField("Visibility_mi", DoubleType(), True),
    StructField("Wind_Direction", StringType(), True),
    StructField("Wind_Speed_mph", DoubleType(), True),
    StructField("Precipitation_in", DoubleType(), True),
    StructField("Weather_Condition", StringType(), True),
    StructField("Amenity", BooleanType(), True),
    StructField("Bump", BooleanType(), True),
    StructField("Crossing", BooleanType(), True),
    StructField("Give_Way", BooleanType(), True),
    StructField("Junction", BooleanType(), True),
    StructField("No_Exit", BooleanType(), True),
    StructField("Railway", BooleanType(), True),
    StructField("Roundabout", BooleanType(), True),
    StructField("Station", BooleanType(), True),
    StructField("Stop", BooleanType(), True),
    StructField("Traffic_Calming", BooleanType(), True),
    StructField("Traffic_Signal", BooleanType(), True),
    StructField("Turning_Loop", BooleanType(), True),
    StructField("Sunrise_Sunset", StringType(), True),
    StructField("Civil_Twilight", StringType(), True),
    StructField("Nautical_Twilight", StringType(), True),
    StructField("Astronomical_Twilight", StringType(), True)
])
```



# consumer.py

```
windowed_df = (  
    parsed_df.withWatermark("Start_Time", "1 hour")  
    .groupBy(windowed_col)  
    .agg(  
        avg("Visibility_mi").alias("Visibility_mi"),  
        avg("Wind_Speed_mph").alias("Wind_Speed_mph"),  
        avg("Temperature_F").alias("Temperature_F"),  
        avg("Humidity_percent").alias("Humidity_percent"),  
        avg("Pressure_in").alias("Pressure_in"),  
        avg("Distance_mi").alias("Distance_mi"),  
        avg("Precipitation_in").alias("Precipitation_in"),  
        avg("Start_Lat").alias("Start_Lat"),  
        avg("Start_Lng").alias("Start_Lng"),  
        avg(col("Junction").cast("double")).alias("Junction"),  
        avg(col("Traffic_Signal").cast("double")).alias("Traffic_Signal"),  
        avg(col("Crossing").cast("double")).alias("Crossing"),  
        mode_udf(collect_list("Weather_Condition")).alias("Weather_Condition"),  
        mode_udf(collect_list("Wind_Direction")).alias("Wind_Direction"),  
        mode_udf(collect_list("Sunrise_Sunset")).alias("Sunrise_Sunset"),  
        count("*").alias("accident_count"),  
    )  
)
```

```
@udf(StringType())  
def mode_udf(values):  
    if not values:  
        return None  
    filtered = [v for v in values if v is not None]  
    if not filtered:  
        return None  
    counter = Counter(filtered)  
    return counter.most_common(1)[0][0]
```

```
windowed_features = windowed_df.select(  
    col("window.start").alias("window_start"),  
    col("window.end").alias("window_end"),  
    "accident_count",  
    "Visibility_mi",  
    "Wind_Speed_mph",  
    "Temperature_F",  
    "Humidity_percent",  
    "Pressure_in",  
    "Distance_mi",  
    "Precipitation_in",  
    "Start_Lat",  
    "Start_Lng",  
    "Junction",  
    "Traffic_Signal",  
    "Crossing",  
    "Weather_Condition",  
    "Wind_Direction",  
    "Sunrise_Sunset"  
)  
  
predictions = model.transform(windowed_features)  
  
predictions.writeStream \  
    .format("json") \  
    .outputMode("append") \  
    .option("path", "output/predictions_json") \  
    .option("checkpointLocation", "output/checkpoint") \  
    .start() \  
    .awaitTermination()
```

# primer json fajla

```
{
  "window_start": "2016-02-12T17:00:00.000+01:00",
  "window_end": "2016-02-12T18:00:00.000+01:00",
  "accident_count": 4,
  "Visibility_mi": 3.0,
  "Wind_Speed_mph": 12.7,
  "Temperature_F": 23.0,
  "Humidity_percent": 80.0,
  "Pressure_in": 30.03,
  "Distance_mi": 1.67,
  "Precipitation_in": 0.0,
  "Start_Lat": 40.109928,
  "Start_Lng": -82.978203,
  "Junction": 1.0,
  "Traffic_Signal": 0.0,
  "Crossing": 0.0,
  "Weather_Condition": "Light Snow",
  "Wind_Direction": "West",
  "Sunrise_Sunset": "Day",
  "Weather_ConditionIndexed": 8.0,
  "Wind_DirectionIndexed": 12.0,
  "Sunrise_SunsetIndexed": 0.0,
  "features": {
    "type": 1,
    "values": [3.0, 12.7, 23.0, 80.0, 30.03, 1.67, 0.0, 40.109928, -82.978203, 1.0, 0.0, 0.0, 8.0, 12.0, 0.0]
  },
  "rawPrediction": {
    "type": 1,
    "values": [-1.077946733725134, 1.077946733725134]
  },
  "probability": {
    "type": 1,
    "values": [0.10378178367571056, 0.8962182163242894]
  },
  "prediction": 1.0
}
```

# Docker compose

```
services:
  kafka:
    image: apache/kafka
    container_name: kafka
    ports:
      - "9092:9092"
    environment:
      KAFKA_NODE_ID: 1
      KAFKA_PROCESS_ROLES: broker,controller
      KAFKA_LISTENERS: PLAINTEXT://0.0.0.0:29092,CONTROLLER://0.0.0.0:9093,PLAINTEXT_HOST://0.0.0.0:9092
      KAFKA_ADVERTISED_LISTENERS: PLAINTEXT://kafka:29092,PLAINTEXT_HOST://localhost:9092
      KAFKA_CONTROLLER_LISTENER_NAMES: CONTROLLER
      KAFKA_LISTENER_SECURITY_PROTOCOL_MAP: CONTROLLER:PLAINTEXT,PLAINTEXT:PLAINTEXT,PLAINTEXT_HOST:PLAINTEXT
      KAFKA_CONTROLLER_QUORUM_VOTERS: 1@localhost:9093
      KAFKA_INTER_BROKER_LISTENER_NAME: PLAINTEXT
      KAFKA_OFFSETS_TOPIC_REPLICATION_FACTOR: 1
      KAFKA_TRANSACTION_STATE_LOG_REPLICATION_FACTOR: 1
      KAFKA_TRANSACTION_STATE_LOG_MIN_ISR: 1
    networks:
      - bigdata-3
```

```
datanode:
  image: bde2020/hadoop-datanode:2.0.0-hadoop3.2.1-java8
  container_name: datanode
  restart: always
  volumes:
    - hadoop_datanode:/hadoop/dfs/data
  environment:
    SERVICE_PRECONDITION: "namenode:9870"
  env_file:
    - ./hadoop.env
  networks:
    - bigdata-3
```

```
resourcemanager:
  image: bde2020/hadoop-resourcemanager:2.0.0-hadoop3.2.1-java8
  container_name: resourcemanager
  restart: always
  environment:
    SERVICE_PRECONDITION: "namenode:9000 namenode:9870 datanode:9864"
  env_file:
    - ./hadoop.env
  networks:
    - bigdata-3
```

```
cassandra:
  image: cassandra:latest
  container_name: cassandra-container
  ports:
    - "9042:9042"
  environment:
    - CASSANDRA_USER=admin
    - CASSANDRA_PASSWORD=admin
    - MAX_HEAP_SIZE=512M
    - HEAP_NEWSIZE=100M
  deploy:
    resources:
      limits:
        memory: 2G
  networks:
    - bigdata-3

namenode:
  image: bde2020/hadoop-namenode:2.0.0-hadoop3.2.1-java8
  container_name: namenode
  restart: always
  ports:
    - "9870:9870"
    - "9000:9000"
  volumes:
    - hadoop_namenode:/hadoop/dfs/name
    - ./models:/hadoop/dfs/models
  environment:
    - CLUSTER_NAME=test
  env_file:
    - ./hadoop.env
  networks:
    - bigdata-3
```

# Docker compose

```
nodemanager:
  image: bde2020/hadoop-nodemanager:2.0.0-hadoop3.2.1-java8
  container_name: nodemanager
  restart: always
  environment:
    SERVICE_PRECONDITION: "namenode:9000 namenode:9870 datanode:9864 resourcemanager:8088"
  env_file:
    - ./hadoop.env
  networks:
    - bigdata-3

historyserver:
  image: bde2020/hadoop-historyserver:2.0.0-hadoop3.2.1-java8
  container_name: historyserver
  restart: always
  environment:
    SERVICE_PRECONDITION: "namenode:9000 namenode:9870 datanode:9864 resourcemanager:8088"
  volumes:
    - hadoop_historyserver:/hadoop/yarn/timeline
  env_file:
    - ./hadoop.env
  networks:
    - bigdata-3

spark-master:
  image: bde2020/spark-master:3.1.2-hadoop3.2
  container_name: spark-master
  ports:
    - "8070:8080"
    - "7077:7077"
  volumes:
    - ./spark-apps:/opt/spark-apps
  environment:
    - INIT_DAEMON_STEP=setup_spark
  networks:
    - bigdata-3
```

```
spark-worker-1:
  image: bde2020/spark-worker:3.1.2-hadoop3.2
  container_name: spark-worker-1
  depends_on:
    - spark-master
  environment:
    - SPARK_MASTER=spark://spark-master:7077
  networks:
    - bigdata-3

spark-worker-2:
  image: bde2020/spark-worker:3.1.2-hadoop3.2
  container_name: spark-worker-2
  depends_on:
    - spark-master
  environment:
    - SPARK_MASTER=spark://spark-master:7077
  networks:
    - bigdata-3

volumes:
  hadoop_namenode:
  hadoop_datanode:
  hadoop_historyserver:

networks:
  bigdata-3:
    driver: bridge
```



# Izgenerisani fajlovi

```
root@12a7cdd362a8:/# hdfs dfs -ls /output/predictions_json
Found 161 items
drwxr-xr-x   - root supergroup          0 2026-01-18 12:22 /output/predictions_json/_spark_metadata
-rw-r--r--   3 root supergroup          0 2026-01-18 12:20 /output/predictions_json/part-00000-04a99b39-c7d1-4467-b130-c11423dd332b-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-072bddb3-f2cc-440c-a5a6-3e7dc6615cbc-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:20 /output/predictions_json/part-00000-0c49d488-1e07-4d6a-a9e3-d07eef29b585-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-0cb464cf-a96a-4fb3-bb6e-8beb4dffc88b-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:17 /output/predictions_json/part-00000-0de0039e-bda1-402b-9f8b-b50894c0ca34-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-0fcb1550-ad66-4609-abe3-2f72bfa305c2-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-11c3fc28-1e89-4e4b-85f5-031e68f71c89-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:16 /output/predictions_json/part-00000-17db4aa4-2aa8-4267-a253-82aad94033c0-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-18a4483e-60db-4ad3-a455-62bdc66b1e2e-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-1bcc0a11-98f3-4e4e-bea9-128e33cb6327-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-1c8f142c-e972-4f93-9c66-8c82ae43793e-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:17 /output/predictions_json/part-00000-1cc9f51e-0023-4ad3-b822-f038b4572070-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-1e713006-2809-4f65-be67-73a951b20fc0-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-20cfd4ab-b34d-43fa-b49f-f907ecad8d63-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:16 /output/predictions_json/part-00000-22637817-56da-4517-8661-dc16caeb74a1-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-230cd6eb-184f-46d4-a695-b50040823e84-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:21 /output/predictions_json/part-00000-2790035d-3bac-4030-b1c9-2ff546526cf7-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-28ca9748-279b-42df-a458-8ecf5d029aa-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:17 /output/predictions_json/part-00000-2a2261a7-a8c0-4db1-97d6-8d234ccc1456-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:17 /output/predictions_json/part-00000-2bb454ed-9eac-42f7-98d7-9d6da9d0540b-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:16 /output/predictions_json/part-00000-3121bbc3-2555-430c-a2ef-2426b47b0ad8-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:21 /output/predictions_json/part-00000-39654b1f-e27d-4375-a676-902bd1c6e505-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-3a06f3e2-b4cd-4e31-92e2-ddd8582643c0-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-3bacd5d3-81e0-4ced-a7d9-23cbba4dcbb0-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-3c330e7f-08b8-491c-bca8-4675d29bb395-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-414204e0-3ab7-46ad-bae2-90eca748fe04-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-416824d6-0ae5-4825-8003-0b52f198bff9-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:16 /output/predictions_json/part-00000-41c06366-6faf-4fde-83db-5e515dc75c0f-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-42eeaab5-8a3c-443e-be70-0453f9391466-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:21 /output/predictions_json/part-00000-43a6cd50-4d6c-42e2-99b4-5a5be76be530-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-48228368-5cbd-49e3-89c7-0f8d809e9874-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:20 /output/predictions_json/part-00000-4a3f8aaf-a8db-4656-b494-d2cd23feafd8-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-4b02ca5b-0d2a-4daf-86f9-6d57e9fc867d-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-4d6739fd-f3ca-4e9b-bbea-f9fb0bc0423b-c000.json
```

```
@echo off
docker exec spark-master sh -c "apk add --no-cache py3-numpy"
docker exec -i namenode bash -c "hdfs dfsadmin -safemode leave"
docker exec namenode hdfs dfs -rm -r /output
docker exec namenode hdfs dfs -mkdir -p /output
docker exec spark-master /spark/bin/spark-submit --master spark://spark-master:7077 --packages org.apache.spark:spark-sql-kafka-0-10_2.12:3.1.2 /opt/spark-apps/consumer.docker.py "tumbling" "1 hour" "1 hour"
pause
```

run. bat



# Izgenerisani fajlovi

```
root@12a7cdd362a8:/# hdfs dfs -ls /output/predictions_json
Found 161 items
drwxr-xr-x   - root supergroup          0 2026-01-18 12:22 /output/predictions_json/_spark_metadata
-rw-r--r--   3 root supergroup          0 2026-01-18 12:20 /output/predictions_json/part-00000-04a99b39-c7d1-4467-b130-c11423dd332b-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-072bddb3-f2cc-440c-a5a6-3e7dc6615cbc-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:20 /output/predictions_json/part-00000-0c49d488-1e07-4d6a-a9e3-d07eef29b585-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-0cb464cf-a96a-4fb3-bb6e-8beb4dfffc88b-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:17 /output/predictions_json/part-00000-0de0039e-bda1-402b-9f8b-b50894c0ca34-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-0fcb1550-ad66-4609-abe3-2f72bfa305c2-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-11c3fc28-1e89-4e4b-85f5-031e68f71c89-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:16 /output/predictions_json/part-00000-17db4aa4-2aa8-4267-a253-82aad94033c0-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-18a4483e-60db-4ad3-a455-62bdc66b1e2e-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-1bcc0a11-98f3-4e4e-bea9-128e33cb6327-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-1c8f142c-e972-4f93-9c66-8c82ae43793e-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:17 /output/predictions_json/part-00000-1cc9f51e-0023-4ad3-b822-f038b4572070-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-1e713006-2809-4f65-be67-73a951b20fc0-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-20cf44ab-b34d-43fa-b49f-f907ecad8d63-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:16 /output/predictions_json/part-00000-22637817-56da-4517-8661-dc16caeb74a1-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-230cd6eb-184f-46d4-a695-b50040823e84-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:21 /output/predictions_json/part-00000-2790035d-3bac-4030-b1c9-2ff546526cf7-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:19 /output/predictions_json/part-00000-28ca9748-279b-42df-a458-8ecf5d029aa-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:17 /output/predictions_json/part-00000-2a2261a7-a8c0-4db1-97d6-8d234ccc1456-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:17 /output/predictions_json/part-00000-2bb454ed-9eac-42f7-98d7-9d6da9d0540b-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:16 /output/predictions_json/part-00000-3121bbc3-2555-430c-a2ef-2426b47b0ad8-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:21 /output/predictions_json/part-00000-39654b1f-e27d-4375-a676-902bd1c6e505-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-3a06f3e2-b4cd-4e31-92e2-ddd8582643c0-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-3bacd5d3-81e0-4ced-a7d9-23cbba4dcbb0-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-3c330e7f-08b8-491c-bca8-4675d29bb395-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-414204e0-3ab7-46ad-bae2-90eca748fe04-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-416824d6-0ae5-4825-8003-0b52f198bff9-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:16 /output/predictions_json/part-00000-41c06366-6faf-4fde-83db-5e515dc75c0f-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-42eeaab5-8a3c-443e-be70-0453f9391466-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:21 /output/predictions_json/part-00000-43a6cd50-4d6c-42e2-99b4-5a5be76be530-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-48228368-5cbd-49e3-89c7-0f8d809e9874-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:20 /output/predictions_json/part-00000-4a3f8aaf-a8db-4656-b494-d2cd23feafd8-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:18 /output/predictions_json/part-00000-4b02ca5b-0d2a-4daf-86f9-6d57e9fc867d-c000.json
-rw-r--r--   3 root supergroup          0 2026-01-18 12:22 /output/predictions_json/part-00000-4d6739fd-f3ca-4e9b-bbea-f9fb0bc0423b-c000.json
```

```
@echo off
docker exec spark-master sh -c "apk add --no-cache py3-numpy"
docker exec -i namenode bash -c "hdfs dfsadmin -safemode leave"
docker exec namenode hdfs dfs -rm -r /output
docker exec namenode hdfs dfs -mkdir -p /output
docker exec spark-master /spark/bin/spark-submit --master spark://spark-master:7077 --packages org.apache.spark:spark-sql-kafka-0-10_2.12:3.1.2 /opt/spark-apps/consumer.docker.py "tumbling" "1 hour" "1 hour"
pause
```

run. bat

# consumer.docker.py

```
spark = SparkSession.builder.appName("Big-Data-3") \
    .getOrCreate()

model = PipelineModel.load("hdfs://namenode:9000/models/accident_gbt_pipeline")
```

```
df = (
    spark.readStream.format("kafka") \
        .option("kafka.bootstrap.servers", "kafka:29092") \
        .option("subscribe", "us_accidents_topic") \
        .option("startingOffsets", "earliest") \
        .load() \
        .selectExpr("CAST(value AS STRING) as csv_line") \
)
```

```
predictions.writeStream \
    .format("json") \
    .outputMode("append") \
    .option("path", "hdfs://namenode:9000/output/predictions_json") \
    .option("checkpointLocation", "hdfs://namenode:9000/output/checkpoint") \
    .start() \
    .awaitTermination()
```



# visualize\_predictions.py

```
import glob
import pandas as pd
import matplotlib.pyplot as plt
import os

def load_predictions(json_folder="output/predictions_json"):
    files = glob.glob(os.path.join(json_folder, "*.json"))

    if not files:
        return None

    df_list = []
    for f in files:
        if os.path.getsize(f) == 0:
            continue
        try:
            tmp = pd.read_json(f, lines=True)
            if not tmp.empty:
                df_list.append(tmp)
        except (ValueError, KeyError):
            continue

    if not df_list:
        return None

    df = pd.concat(df_list, ignore_index=True)
    df["window_start"] = pd.to_datetime(df["window_start"])
    df["high_severity_prob"] = df["probability"].apply(lambda x: x["values"][1])
    df = df.sort_values("window_start")

    return df
```

```
def plot_probability_histogram(df):
    fig, ax = plt.subplots(figsize=(10, 5))

    ax.hist(df["high_severity_prob"], bins=20, color="steelblue", edgecolor="black")
    ax.axvline(x=0.5, color="red", linestyle="--", label="Prag odluke (0.5)")
    ax.set_title("Histogram verovatnoće visoke ozbiljnosti nesreća")
    ax.set_xlabel("Verovatnoća visoke ozbiljnosti")
    ax.set_ylabel("Broj prozora")
    ax.legend()

    plt.tight_layout()
    plt.savefig("output/probability_histogram.png", dpi=150)
    plt.show()
```

```
def plot_weather_analysis(df):
    weather_stats = df.groupby("Weather_Condition").agg({
        "prediction": "mean",
        "high_severity_prob": "mean",
        "accident_count": "sum"
    }).round(3)

    weather_stats.columns = ["Avg_Prediction", "Avg_Probability", "Total_Accidents"]
    weather_stats = weather_stats[weather_stats.index != "Clear"]
    weather_stats = weather_stats.sort_values("Avg_Probability", ascending=False).head(15)

    fig, ax = plt.subplots(figsize=(12, 6))

    ax.barh(weather_stats.index, weather_stats["Avg_Probability"], color="coral")
    ax.axvline(x=0.5, color="black", linestyle="--", alpha=0.5)
    ax.set_xlabel("Prosečna verovatnoća visoke ozbiljnosti")
    ax.set_title("Vremenski uslovi po ozbiljnosti nesreća")

    plt.tight_layout()
    plt.savefig("output/weather_risk_analysis.png", dpi=150)
    plt.show()
```

# visualize\_predictions.py

```
def plot_hourly_weekday_heatmap(df):

    df_copy = df.copy()
    df_copy["hour"] = df_copy["window_start"].dt.hour
    df_copy["weekday"] = df_copy["window_start"].dt.dayofweek

    pivot = df_copy.pivot_table(
        values="high_severity_prob",
        index="hour",
        columns="weekday",
        aggfunc="mean"
    )

    day_names = ["Pon", "Uto", "Sre", "Čet", "Pet", "Sub", "Ned"]
    pivot.columns = [day_names[i] for i in pivot.columns]

    fig, ax = plt.subplots(figsize=(10, 8))

    im = ax.imshow(pivot.values, cmap="YlOrRd", aspect="auto")

    ax.set_xticks(range(len(pivot.columns)))
    ax.set_xticklabels(pivot.columns)
    ax.set_yticks(range(len(pivot.index)))
    ax.set_yticklabels([f"{h:02d}:00" for h in pivot.index])

    cbar = plt.colorbar(im, ax=ax)
    cbar.set_label("Prosečna verovatnoća visoke ozbiljnosti")

    ax.set_xlabel("Dan u nedelji")
    ax.set_ylabel("Sat")
    ax.set_title("Rizik ozbiljnih nesreća po satu i danu u nedelji")

    plt.tight_layout()
    plt.savefig("output/hourly_weekday_heatmap.png", dpi=150)
    plt.show()

if __name__ == "__main__":
    os.makedirs("output", exist_ok=True)

    df = load_predictions()

    if df is not None and not df.empty:
        plot_probability_histogram(df)
        plot_weather_analysis(df)
        plot_hourly_weekday_heatmap(df)
```

# visualize\_predictions.py

```
def plot_hourly_weekday_heatmap(df):

    df_copy = df.copy()
    df_copy["hour"] = df_copy["window_start"].dt.hour
    df_copy["weekday"] = df_copy["window_start"].dt.dayofweek

    pivot = df_copy.pivot_table(
        values="high_severity_prob",
        index="hour",
        columns="weekday",
        aggfunc="mean"
    )

    day_names = ["Pon", "Uto", "Sre", "Čet", "Pet", "Sub", "Ned"]
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    cbar = plt.colorbar(im, ax=ax)
    cbar.set_label("Prosečna verovatnoća visoke ozbiljnosti")

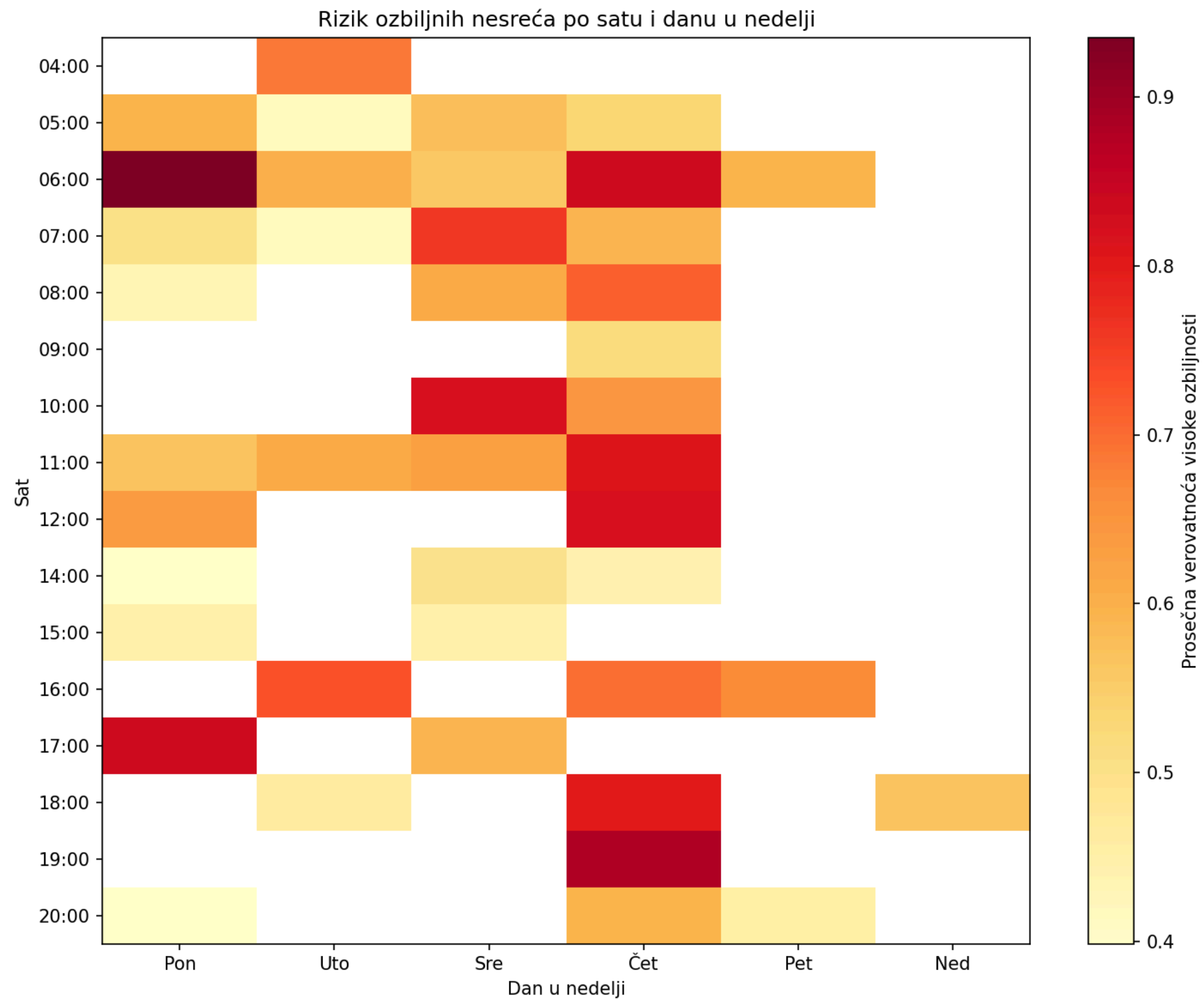
    ax.set_xlabel("Dan u nedelji")
    ax.set_ylabel("Sat")
    ax.set_title("Rizik ozbiljnih nesreća po satu i danu u nedelji")

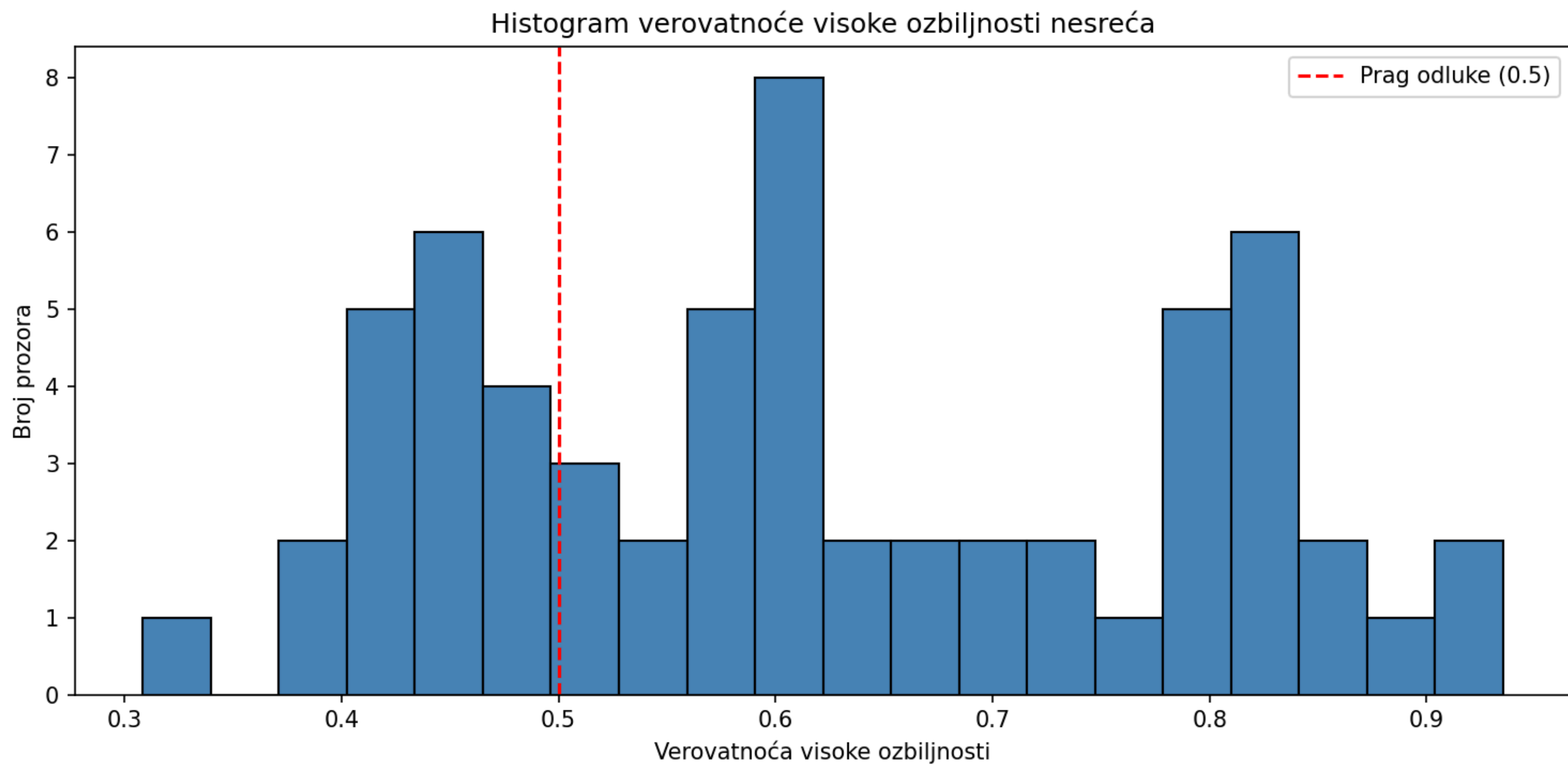
    plt.tight_layout()
    plt.savefig("output/hourly_weekday_heatmap.png", dpi=150)
    plt.show()

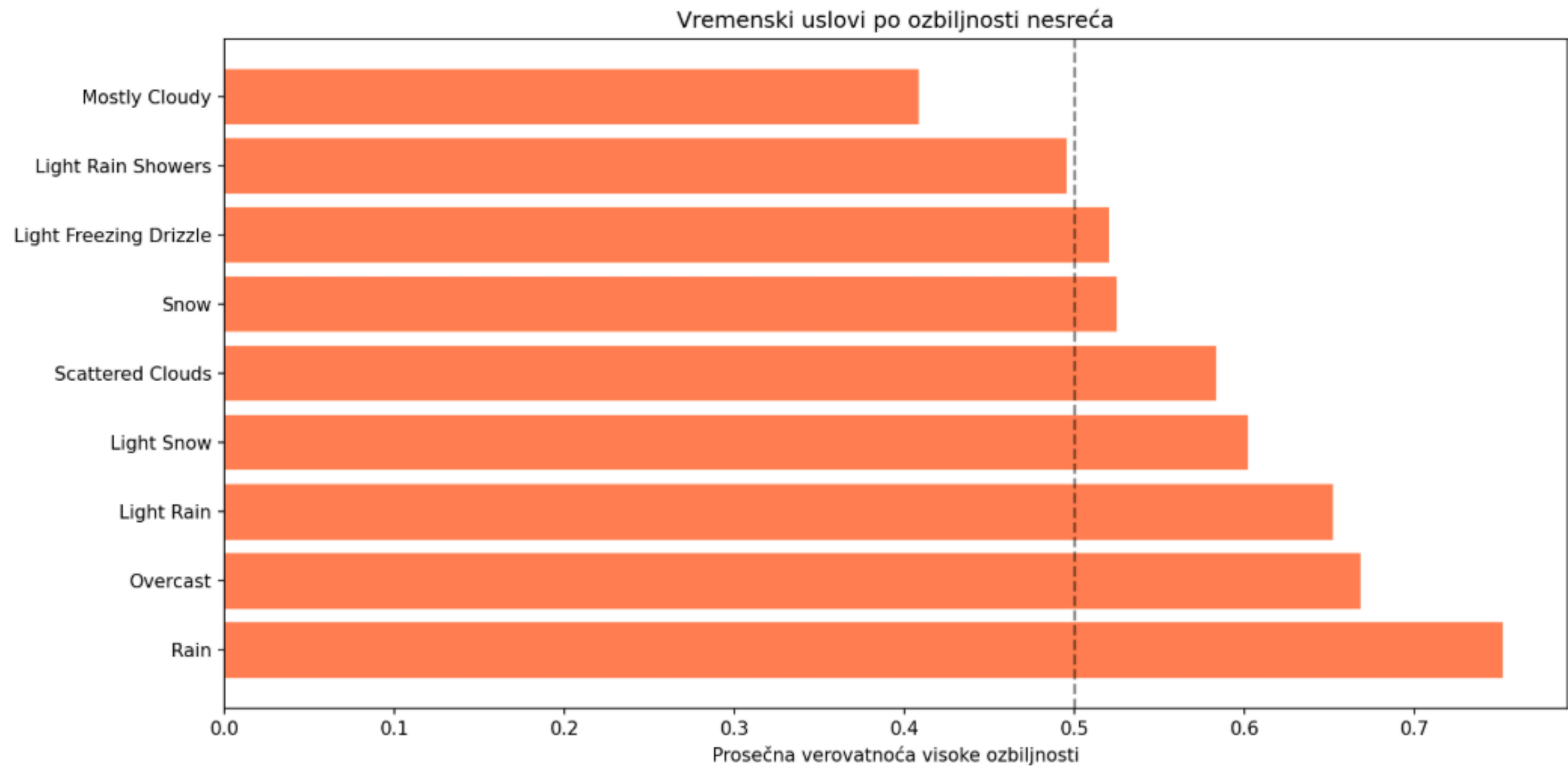
if __name__ == "__main__":
    os.makedirs("output", exist_ok=True)

    df = load_predictions()

    if df is not None and not df.empty:
        plot_probability_histogram(df)
        plot_weather_analysis(df)
        plot_hourly_weekday_heatmap(df)
```







**Hvala na pažnji!**