# Projekat 3 — Big data

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# Kreiranje modela (app\_3a.py)

#### Dodavanje klasnog atributa "visit\_in\_worktime"

```
# Kreiranje kolone koja sadrži sat cekiranja
df_hour = df.withColumn('visit_hour', hour('check_in_time'))

# Kreiranje kolone 'visit_in_worktime' koja će imati vrednost 'True' ako je cekiranje u radno vreme, a 'False' ako nije
df_worktime = df_hour.withColumn('visit_in_worktime', when((col('visit_hour') >= 9) & (col('visit_hour') < 17), True).otherwise(False))
df_worktime = df_worktime.drop('visit_hour', 'time_spent', 'location_id')</pre>
```

- Dodaje se klasni atribut "visit\_in\_worktime" koji pokazauje da li je čekiranje za vreme radnog vremena ili ne
- Atrbiut uzima vrednost True ukoliko je vreme čekiranja između 9č i 17č, u suprotnom je njegova vrednost False

# Priprema i obučavanje modela

```
# Podela na training i test set
train, test = df worktime.randomSplit([0.7, 0.3], seed=42)
# Konvertovanje kolone 'visit in worktime' u 0 ili 1
train = train.withColumn('visit in worktime', when(col('visit in worktime') == True, 1).otherwise(0))
test = test.withColumn('visit in worktime', when(col('visit in worktime') == True, 1).otherwise(0))
# Kreiranje vektora feature-a
feature cols = ["user", "check in time", "latitude", "longitude"]
assembler = VectorAssembler(inputCols=feature cols, outputCol="features")
# Transformacija skupa za treniranje i skup za testiranje
train = assembler.transform(train).select("features", "visit in worktime")
test = assembler.transform(test).select("features", "visit in worktime")
# Kreiranje i treniranje Decision Tree klasifikatora
dt = DecisionTreeClassifier(featuresCol="features", labelCol="visit in worktime")
model = dt.fit(train)
model.save(sys.argv[2])
```

- Vrši se podela na test i trening skup podataka
- Biraju se kolone koje će se koristiti za obučavanje modela
- Vrši se obučavanje modela
- Model se čuva na HDFS-u

# Evaluacija modela

```
# Klasifikacija test set-a
predictions = model.transform(test)
# Evaluacija klasifikatora
evaluator = MulticlassClassificationEvaluator(predictionCol="prediction", labelCol="visit in worktime", metricName="accuracy")
accuracy = evaluator.evaluate(predictions)
print("Tacnost modela: ", accuracy)
f1 evaluator = MulticlassClassificationEvaluator(predictionCol="prediction", labelCol="visit in worktime", metricName="f1")
f1 score = f1 evaluator.evaluate(predictions)
print("F1 skor modela: ", f1 score)
# Recall
recall evaluator = MulticlassClassificationEvaluator(predictionCol="prediction", labelCol="visit in worktime", metricName="weightedRecall")
recall = recall evaluator.evaluate(predictions)
print("Odziv modela: ", recall)
# Precision
precision evaluator = MulticlassClassificationEvaluator(predictionCol="prediction", labelCol="visit in worktime", metricName="weightedPrecision")
precision = precision evaluator.evaluate(predictions)
print("Preciznost modela: ", precision)
print("Raspodela po klasama")
predictions.groupBy("visit_in_worktime").count().show()
# Matrica konfuzije
predictions = predictions.withColumn("visit in worktime", col("visit in worktime").cast("double"))
predictionAndLabels = predictions.select("prediction", "visit in worktime").rdd
metrics = MulticlassMetrics(predictionAndLabels)
print("Matrica konfuzije:")
print(metrics.confusionMatrix().toArray())
```

- Evaluacija modela se vrši nad test skupom podataka
- Kao mere za evaluaciju se koriste tačnost, f1 – mera, odziv, preciznost i matrica konfuzije

## Evaluacija modela - rezultati

```
Tacnost modela: 0.9019337016574586
F1 skor modela: 0.9017181005939445
Odziv modela: 0.9019337016574586
Preciznost modela: 0.9067808826673667
Raspodela po klasama
+----+
|visit_in_worktime|count|
               1 1467
Matrica konfuzije:
[[1365. 64.]
 [ 220. 1247.]]
```

# Skripta za pokretanje aplikacije

#!/bin/bash

spark/bin/spark-submit --master spark://spark-master:7077 app\_3.py hdfs://namenode:9000/data\_3.csv hdfs://namenode:9000/DecisionTreeModel\_12

- Kao drugi argument navodimo putanju do skupa podataka koji će se koristiti za obučavanje i evaluaciju modela, dok je treći argument putanja do lokacije na HDFS-u gde želimo snimiti model
- Za obučavnje modela je iskorišćen data\_3.csv skup podataka, koji sadrži 10000 slogova, i koji je kreiran na osnovu originalnog brightkite skupa podataka
- U data\_3.csv se nalazi po 5000 instanci iz obe klase, da bi se uspostavila balansiranost



#### Spork 3.1.2 Application: Project\_3A

ID: app-20230322133408-0003

Name: Project\_3A User: root

Cores: Unlimited (16 granted) Executor Limit: Unlimited (2 granted)

Executor Memory: 1024.0 MiB

Executor Resources:

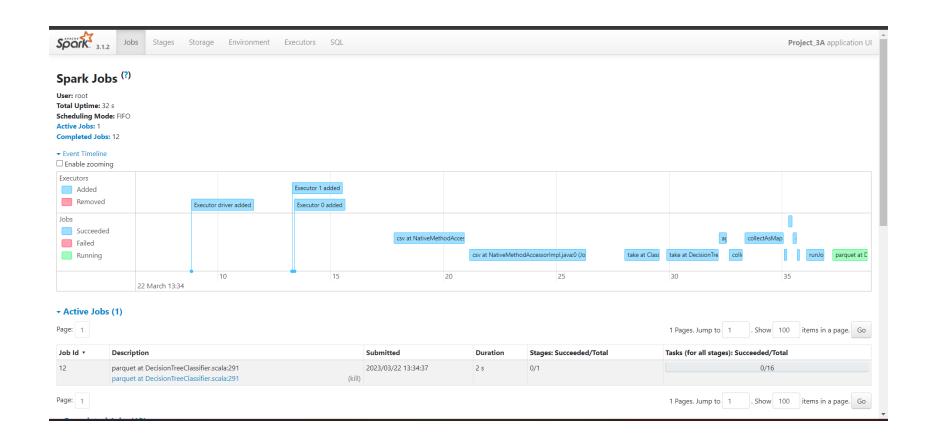
Submit Date: 2023/03/22 13:34:08

State: RUNNING

Application Detail UI

#### ▼ Executor Summary (2)

ExecutorID	Worker	Cores	Memory	Resources	State	Logs
1	worker-20230322125054-172.23.0.5-40587	8	1024		RUNNING	stdout stderr
0	worker-20230322125055-172.18.0.6-38403	8	1024		RUNNING	stdout stderr



# Klasifikacija podataka sa Kafka topic-a (app\_3b.py)

# Čitanje i parsiranje podataka sa Kafka topic-a

```
df = spark \
  .readStream \
  .format("kafka") \
  .option("kafka.bootstrap.servers", "kafka:9092") \
  .option("subscribe", "topic2") \
  .option("startingOffsets", "earliest") \
  .option("maxOffsetsPerTrigger", 1) \
  .load()
parsed df = df.selectExpr("CAST(value AS STRING)")
split col = split(parsed df['value'], ',')
parsed df = parsed df.withColumn("user", split col.getItem(0))
parsed df = parsed df.withColumn("check in time", split col.getItem(1))
parsed df = parsed df.withColumn("latitude", split col.getItem(2))
parsed df = parsed df.withColumn("longitude", split col.getItem(3))
parsed df = parsed df.withColumn("location id", split col.getItem(4))
parsed df = parsed df.withColumn("time spent", split col.getItem(5))
parsed df = parsed df.withColumn("time stamp", split col.getItem(6))
parsed df = parsed df.where(parsed df.user != 'user')
parsed df = parsed df.withColumn("check in time", to timestamp(parsed df["check in time"], "yyyy-MM-dd'T'HH:mm:ss.SSSXXX"))
parsed df = parsed df.withColumn('check in time', unix timestamp('check in time'))
parsed df = parsed df.withColumn("user", parsed df["user"].cast(IntegerType()))
parsed_df = parsed_df.withColumn("latitude", parsed_df["latitude"].cast(DoubleType()))
parsed df = parsed df.withColumn("longitude", parsed df["longitude"].cast(DoubleType()))
```

# Učitavanje modela i predikcije za podatke sa Kafka topic-a

```
# Ucitavanje modela
model_path = "hdfs://namenode:9000/DecisionTreeModel_5"
dt_model = DecisionTreeClassificationModel.load(model_path)

assembler = VectorAssembler(
    inputCols=["user", "check_in_time", "latitude", "longitude"],
    outputCol="features")

parsed_df = assembler.transform(parsed_df)

# Klasifikacija
predictions = dt_model.transform(parsed_df)
```

## Upis rezultata u InfluxDB

```
def write to influxdb(df, epoch id):
    # Inicijalizacija InfluxDB klijenta
    token = "X199hGe7tyPW-wWrgpZRo8vOxA6bK2nR-X3MEoqkigZqnSG1vSqpKOoBmLZdWpWbYKKMKNEfqAAX4FMoKhd5ug=="
   org = "brightkite-org"
    bucket = "brightkite-bucket"
    client = InfluxDBClient(url="http://influxdb:8086", token=token)
    # Kreiranje instance WriteApi klase
   write api = client.write api(write options=SYNCHRONOUS)
    for row in df.collect():
        point = Point("predictions 5b") \
            .field("user", row.user) \
            .field("latitude", row.latitude) \
            .field("longitude", row.longitude) \
            .field("visit in worktime", row.prediction)
       write_api.write(bucket=bucket, org=org, record=point)
```

# Skripta za pokretanje

#!/bin/bash

spark/bin/spark-submit --master spark://spark-master:7077 --packages org.apache.spark:spark-sql-kafka-0-10\_2.12:3.1.2 app\_4.py hdfs://namenode:9000/DecisionTreeModel\_5

- Kao drugi argument komande linije navodimo putanju do modela na HDFS-u



#### Spark <sub>3.1.2</sub> Application: Project\_3B

ID: app-20230322134051-0005 Name: Project\_3B

User: root

Cores: Unlimited (16 granted)

Executor Limit: Unlimited (2 granted) Executor Memory: 1024.0 MiB

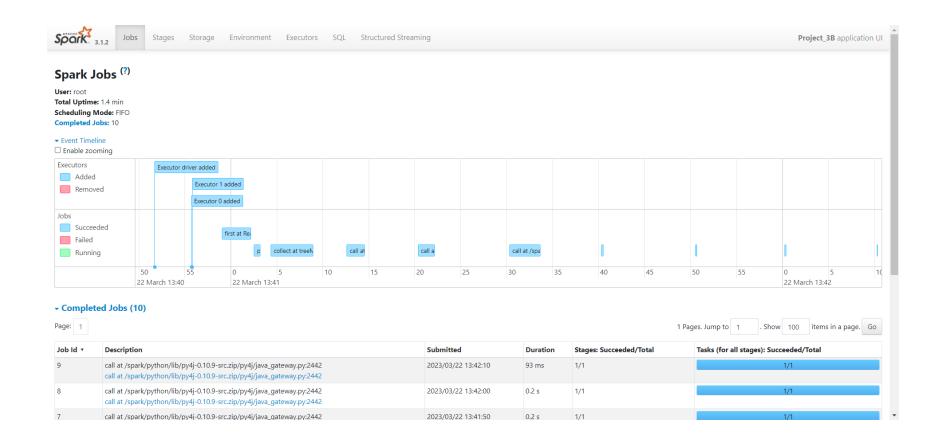
Executor Resources:

Submit Date: 2023/03/22 13:40:51

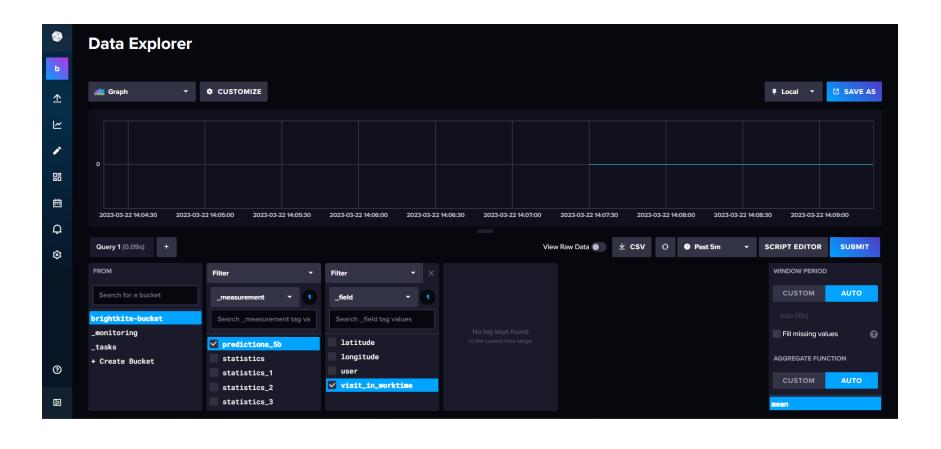
State: RUNNING **Application Detail UI** 

#### **▼** Executor Summary (2)

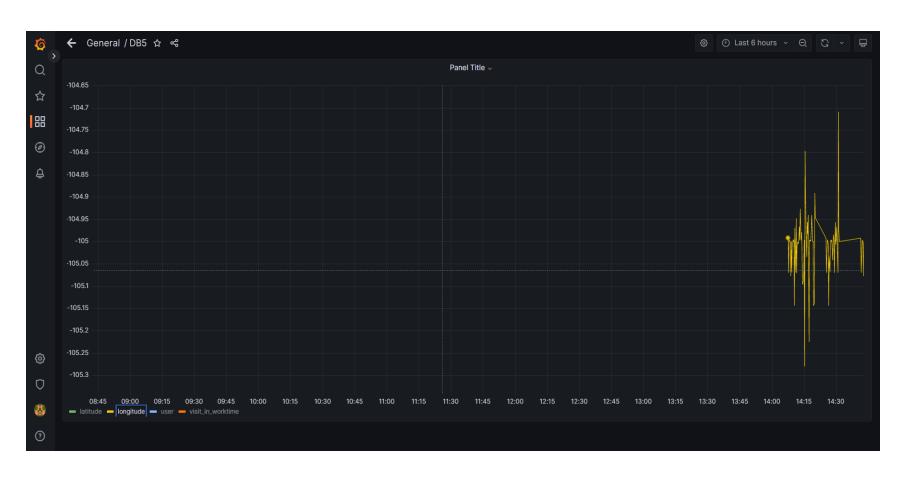
ExecutorID	Worker	Cores	Memory	Resources	State	Logs
1	worker-20230322125054-172.23.0.5-40587	8	1024		RUNNING	stdout stderr
0	worker-20230322125055-172.18.0.6-38403	8	1024		RUNNING	stdout stderr



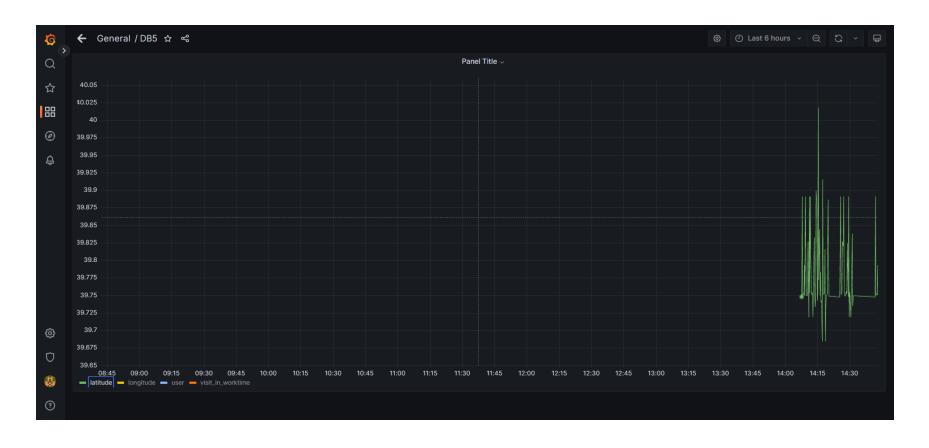
#### Rezultat u InfluxDB



# Vizuelizacija rezultata u Grafani



# Vizuelizacija rezultata u Grafani



# Vizuelizacija rezultata u Grafani

