LAPORAN UJIAN TENGAH SEMESTER MATA KULIAH BIG DATA Jobsheet 10



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PROGRAM STUDI D4 TEKNIK INFORMATIKA JURUSAN TEKNOLOGI INFORMASI POLITEKNIK NEGERI MALANG 2025

Persiapan

```
(base) PS D:\polinema\tugas_kuliah\semester6\big_Data> docker cp ecommerce_transactions_1000.csv spark-ma ster:/opt/spark_data/
Successfully copied 58.4kB to spark-master:/opt/spark_data/
(base) PS D:\polinema\tugas_kuliah\semester6\big_Data>

(base) PS C:\users\user> docker exec -u root -it spark-master bash
root@e9e019bcf75c:/opt/spark/work-dir# ls /opt/spark_data/
ecommerce_transactions_1000.csv
```

Praktikum 1

1. Load Data

2. Inspeksi Data

```
[2]:
  #Inspeksi Data
  #Lihat sturktur schema
  df.printSchema()
 root
  |-- transaction_id: string (nullable = true)
  |-- user_id: string (nullable = true)
  |-- amount: double (nullable = true)
  |-- email: string (nullable = true)
  |-- transaction_time: timestamp (nullable = true)
[3]:
#Hitung missing values setiap kolom
from pyspark.sql.functions import col, when, count
df.select([count(when(col(c).isNull(), c)).alias(c) for c in df.columns]).show()
+----+
|transaction_id|user_id|amount|email|transaction_time|
+----
   0 0 316 0
+----+
```

```
#Hitung jumlah total data
print("Jumlah baris:", df.count())
```

Jumlah baris: 1000

3. Cleaning Data

```
#Cleaning data
 df = df.dropna(subset=["transaction_time"])
 df = df.fillna({"amount":0})
 \label{eq:df.select([count(when(col(c).isNull(), c)).alias(c) for c in df.columns]).show()} \\
 +----
 |transaction_id|user_id|amount|email|transaction_time|
 +-----
          0 0 0 0 0
 +----+
: # Cleaning format email
 from pyspark.sql.functions import instr, substring_index
 #Tambah kolom email_domain
 df = df.withColumn("email_domain", substring_index("email","@", -1))
 #Filter hanya email yang mengandung '0'
 df = df.filter(instr(col("email"),"@")>0)
from pyspark.sql.functions import col, when, count
 df.select(
       when(~col("email").contains("@"), True)
    ).alias("invalid_email_count")
 ).show()
  |invalid_email_count|
  +----+
```

4. Transformasi Data

```
#Transformasi data

from pyspark.sql.types import DoubleType
from pyspark.sql.functions import to_date

df = df.withColumn("amount", col("amount").cast(DoubleType()))
df = df.withColumn("transaction_date", to_date("transaction_time"))
```

5. Simpan Data Bersih

```
#Simpan data bersih
df.write.csv("cleaned_transactions_1000.csv", header =True, mode="overwrite")
```

Pertanyaan & Jawaban

1. Berapa banyak data yang dibuang karena transaction_time kosong?

Sebelum cleaning data

```
#Hitung missing values setiap kolom

from pyspark.sql.functions import col, when, count

df.select([count(when(col(c).isNull(), c)).alias(c) for c in df.columns]).show()

transaction_id|user_id|amount|email|transaction_time|

0 0 316 0 50
```

Proses cleaning data

```
df.filter(df["transaction_time"].isNull())
```

[34]:

DataFrame[transaction_id: string, user_id: string, amount: double, email: string, transaction_time: timestamp, email_domain: string, transaction_date: date]

Sesudah cleaning data

50 data transaction_time berhasil dibuang

2. Apakah semua data amount sudah bertipe numerik setelah cleaning?

Iya, semua data telah bernilai numerik

3. Kenapa lebih baik memperbaiki email invalid sebelum menganalisis data transaksi?

Dikarenakan, email berguna sebagai identifikasi pengguna. Sehinga apabila terdapat masalah invalid email dapat menghasilkan analisis yang salah

Praktikum 2

1. Load Data

```
#Load Data
from pyspark.sql import SparkSession

spark = SparkSession.builder.appName("OutlierDetection").getOrCreate()

df = spark.read.csv("ecommerce_transactions_1000.csv", header=True, inferSchema=True)

df = df.withColumn("amount", df["amount"].cast("double"))
```

2. Hitung Statistik Dasar

```
quantiles = df.approxQuantile("amount", [0.25, 0.75], 0.05)
Q1, Q3 = quantiles
IQR = Q3 - Q1

lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR

print(f"Q1 = {Q1}, Q3 = {Q3}, IQR = {IQR}")
print(f"Lower Bound = {lower_bound}, Upper Bound = {upper_bound}")
Q1 = 34005.04, Q3 = 74468.55, IQR = 40463.51
Lower Bound = -26690.225, Upper Bound = 135163.815
```

3. Deteksi Outliers

```
outliers = df.filter((df.amount < lower_bound) | (df.amount > upper_bound))
outliers.show()
                                     _id|amount| email| transaction_time|
|transaction_id|user_id|amount|
                 T0008 U212 
T0010 U033
                                                NaN| dgreen@hotmail.com|2025-04-23 07:19:12|
                                                NaN rebecca69@hotmail... 2025-04-15 04:04:31
                 T0013
T0014
                                U184
U130
                                                NaN|jackielewis@yahoo...|2025-03-29 21:00:47|
NaN| dawn56@roman.net|2025-04-15 19:21:50|
                                               NaN | dawn56@roman.net | 2025-04-15 19:21:50|
NaN | hgarcia@yahoo.com | 2025-04-12 00:43:15|
NaN | paul68@yahoo.com | 2025-04-15 11:48:24|
NaN | ysi1va@gmail.com | 2025-04-05 14:14:18|
NaN | shawn41@yahoo.com | 2025-04-26 23:15:02|
NaN | davidsalinas | 2025-04-26 14:43:19|
                  T0019
                                 U280 İ
                                U057|
U157|
U085|
U126|
                  T0020
                  T0022
                 T0023
T0025
                                               NaN| edizabethmclean@p...2025-04-09 15:47:48;
NaN|elizabethmclean@p...2025-04-26 14:43:19|
NaN|taylorjoseph@hotm...|2025-04-16 07:45:18|
NaN| debra62@mail.com|2025-04-20 04:48:33|
NaN| bownanryan@gmail.com|2025-04-23 11:25:08|
NaN|smithdanny@yahoo.com|2025-04-13 11:13:00|
                  T0028
                                 U110
                 T0032
T0033
                                 U060
                 T0039 |
T0040 |
                                U124
U200
                  T0045
                                                NaN|garciajenny@crosb...|2025-04-20 00:46:25|
NaN|michaelaramos@yah...|2025-04-13 12:13:05|
                                 U245
                                 U051
                                                NaN
                                                              ubrown@reves.com|2025-04-03 16:07:33|
                                                NaN| ubrown@reyes.com|2025-04-03 16:07:33|
NaN|michellehale@yaho...|2025-04-22 08:05:29|
NaN|michael35@hotmail...|2025-04-17 21:55:07|
                  T0055
                                 U181
                 T0064
                                 U295
                                                NaN andrea13@gallegos... 2025-04-20 11:05:49
only showing top 20 rows
```

4. Hitung Berapa Banyak Outliers

```
[44]:
print("Jumlah Outliers: ", outliers.count())
Jumlah Outliers: 331
```

Tugas Praktikum

1. Tampilkan top 5 transaksi dengan amount terbesar?

```
# Tugas Praktikum
# Nomer 1
df.orderBy(df["amount"].desc()).show(5)
```

```
|transaction_id|user_id|amount| email transaction_time|
         T0019 | U280 | NaN | hgarcia@yahoo.com | 2025-04-12 00:43:15 |
         T0028 | U110 | NaN elizabethmclean@p... | 2025-04-26 14:43:19 |
         T0020| U057| NaN| pau168@yahoo.com|2025-04-15 11:48:24|
T0014| U130| NaN| dawn56@roman.net|2025-04-15 19:21:50|
T0022| U157| NaN| ysilva@gmail.com|2025-04-05 14:14:18|
+-----
only showing top 5 rows
```

2. Hitung jumlah total transaksi?

```
total transaksi = df.count()
print("Jumlah transaksi:", total_transaksi)
Jumlah transaksi: 1000
```

3. Hitung jumlah outlier?

```
q1, q3 = df.approxQuantile("amount", [0.25, 0.75], 0.01)
iqr = q3 - q1
lower_bound = q1 - 1.5 * iqr
upper_bound = q3 + 1.5 * iqr
outliers = df.filter((df["amount"] < lower_bound) | (df["amount"] > upper_bound))
jumlah_outlier = outliers.count()
print("Jumlah outlier:", jumlah_outlier)
```

Jumlah outlier: 331

4. Hitung persentase outlier terhadap seluruh transaksi?

```
persentase = (jumlah_outlier / total_transaksi) * 100
print(f"Persentase outlier: {persentase:.2f}%")
```

Persentase outlier: 33.10%