
USE CASE

Ride-Sharing Trips Analytics (Uber / Ola-style)

This prepares cohorts for aggregations, performance analysis, and later window functions.

DATASET: Ride Trips (45 Records)

Schema

- trip_id
 - rider_name
 - city
 - driver_name
 - vehicle_type
 - distance_km
 - trip_fare
 - trip_duration_minutes
 - payment_mode
 - trip_status
-

Create the Dataset in PySpark

```
data = [  
    ("T001", "Amit", "Hyderabad", "Ramesh", "Sedan", 12.5, 320, 28, "UPI", "Comple  
    ("T002", "Neha", "Bangalore", "Suresh", "Mini", 8.2, 210, 22, "Card", "Comple  
    ("T003", "Rahul", "Delhi", "Anil", "Bike", 5.1, 120, 15, "Cash", "Completed"),  
    ("T004", "Pooja", "Mumbai", "Vikas", "SUV", 18.0, 560, 45, "UPI", "Cancelled")  
    ("T005", "Arjun", "Chennai", "Kumar", "Mini", 7.8, 200, 20, "UPI", "Completed"  
    ("T006", "Sneha", "Hyderabad", "Ramesh", "Sedan", 14.2, 360, 32, "Card", "Comp  
    ("T007", "Karan", "Delhi", "Anil", "Bike", 6.3, 140, 18, "UPI", "Completed"),  
    ("T008", "Riya", "Bangalore", "Suresh", "Sedan", 11.0, 300, 27, "Wallet", "Con  
    ("T009", "Vikas", "Mumbai", "Vijay", "SUV", 20.5, 650, 50, "Card", "Completed"
```

```
(
    ("T010", "Anjali", "Chennai", "Kumar", "Bike", 4.9, 110, 14, "Cash", "Complete"),
    ("T011", "Farhan", "Delhi", "Anil", "Mini", 9.6, 240, 25, "UPI", "Completed"),
    ("T012", "Megha", "Hyderabad", "Ramesh", "SUV", 19.2, 610, 48, "Card", "Cancel"),
    ("T013", "Suresh", "Bangalore", "Suresh", "Sedan", 13.0, 340, 30, "UPI", "Comple"),
    ("T014", "Divya", "Mumbai", "Vikas", "Mini", 10.2, 260, 26, "Wallet", "Comple"),
    ("T015", "Nikhil", "Delhi", "Anil", "Sedan", 15.5, 390, 34, "UPI", "Completed"),
    ("T016", "Kavya", "Chennai", "Kumar", "Sedan", 12.1, 315, 29, "UPI", "Complete"),
    ("T017", "Rohit", "Hyderabad", "Ramesh", "SUV", 22.0, 700, 55, "Card", "Comple"),
    ("T018", "Simran", "Bangalore", "Suresh", "Bike", 5.8, 130, 16, "Cash", "Comple"),
    ("T019", "Ayesha", "Mumbai", "Vijay", "Mini", 9.9, 250, 24, "UPI", "Completed"),
    ("T020", "Manish", "Delhi", "Anil", "Bike", 6.0, 135, 17, "Wallet", "Completec"),
    ("T021", "Priya", "Hyderabad", "Ramesh", "Sedan", 14.8, 380, 33, "Card", "Comp"),
    ("T022", "Yash", "Chennai", "Kumar", "SUV", 21.3, 680, 52, "UPI", "Completed"),
    ("T023", "Naina", "Bangalore", "Suresh", "Mini", 10.7, 270, 28, "UPI", "Comple"),
    ("T024", "Sameer", "Mumbai", "Vikas", "Sedan", 13.9, 350, 31, "Wallet", "Comple"),
    ("T025", "Ritika", "Delhi", "Anil", "Bike", 5.4, 125, 16, "Cash", "Completed"),
    ("T026", "Gopal", "Hyderabad", "Ramesh", "Mini", 8.9, 225, 23, "UPI", "Comple"),
    ("T027", "Tina", "Bangalore", "Suresh", "Sedan", 12.6, 330, 29, "Card", "Comple"),
    ("T028", "Irfan", "Mumbai", "Vijay", "SUV", 23.4, 740, 58, "Card", "Completed"),
    ("T029", "Sahil", "Chennai", "Kumar", "Mini", 9.4, 235, 24, "UPI", "Completed"),
    ("T030", "Lavanya", "Delhi", "Anil", "Sedan", 14.1, 365, 32, "Wallet", "Comple"),
    ("T031", "Deepak", "Hyderabad", "Ramesh", "Bike", 6.7, 150, 18, "Cash", "Comple"),
    ("T032", "Shweta", "Bangalore", "Suresh", "Mini", 10.0, 255, 26, "UPI", "Comple"),
    ("T033", "Aman", "Mumbai", "Vikas", "Sedan", 15.8, 395, 35, "Card", "Completec"),
    ("T034", "Rekha", "Chennai", "Kumar", "Sedan", 13.5, 345, 30, "UPI", "Complete"),
    ("T035", "Zubin", "Delhi", "Anil", "SUV", 24.0, 760, 60, "Card", "Completed"),
    ("T036", "Pallavi", "Hyderabad", "Ramesh", "Mini", 9.1, 230, 23, "Wallet", "Cc"),
    ("T037", "Naveen", "Bangalore", "Suresh", "Bike", 5.9, 135, 17, "UPI", "Comple"),
    ("T038", "Sonia", "Mumbai", "Vijay", "SUV", 21.7, 690, 54, "Card", "Completed"),
    ("T039", "Harish", "Chennai", "Kumar", "Mini", 8.5, 215, 21, "Cash", "Complete"),
    ("T040", "Kriti", "Delhi", "Anil", "Sedan", 14.6, 375, 33, "UPI", "Completed"),
    ("T041", "Apoorva", "Hyderabad", "Ramesh", "Sedan", 13.2, 335, 30, "Card", "Cc"),
    ("T042", "Mohit", "Bangalore", "Suresh", "SUV", 19.9, 620, 49, "UPI", "Comple"),
    ("T043", "Tanvi", "Mumbai", "Vikas", "Mini", 10.4, 265, 27, "Wallet", "Comple"),
    ("T044", "Rakesh", "Chennai", "Kumar", "Bike", 6.2, 140, 18, "Cash", "Complete"),
    ("T045", "Isha", "Delhi", "Anil", "Mini", 9.7, 245, 25, "UPI", "Completed")
)
```

```
]
```

```
columns = [
    "trip_id", "rider_name", "city", "driver_name", "vehicle_type",
    "distance_km", "trip_fare", "trip_duration_minutes",
    "payment_mode", "trip_status"
```

```
]
```

```
df = spark.createDataFrame(data, columns)
df.show()
df.printSchema()
```

EXERCISES – MEDIUM LEVEL

CSV, JSON, PARQUET (Ride-Sharing Use Case)

SECTION A – CSV

Exercise 1

Write the full dataset to CSV with header enabled.

Output:

```
trips_csv/
```

Exercise 2

Read the CSV and filter:

- `trip_fare > 400`
 - `trip_status = "Completed"`
-

Exercise 3

From CSV, select:

- `trip_id`
- `city`
- `vehicle_type`
- `trip_fare`

Sort by `trip_fare` descending.

Exercise 4

Write only Bike trips to CSV using delimiter `|`.

SECTION B – JSON

Exercise 5

Write only trips from Mumbai to JSON.

Output:

```
mumbai_trips_json/
```

Exercise 6

Read JSON and add a column:

```
fare_per_km = trip_fare / distance_km
```

Write back to JSON.

Exercise 7

Filter JSON data:

- `payment_mode = "Card"`
 - `vehicle_type = "SUV"`
-

Exercise 8

Force JSON output into a single partition and observe the output structure.

SECTION C – PARQUET

Exercise 9

Convert full dataset to Parquet.

Output:

```
trips_parquet/
```

Exercise 10

Read Parquet and filter:

- `trip_duration_minutes > 45`
-

Exercise 11

Sort Parquet data by `distance_km` descending and write top 10 trips back to Parquet.

Exercise 12

Compare storage size of:

- CSV
- JSON
- Parquet

Answer which is smallest and why.

SECTION D – FORMAT CONVERSION

Exercise 13

Convert:

- CSV → Parquet
 - JSON → Parquet
-

Exercise 14

Read Parquet and write it back as CSV with header and delimiter , .

ANALYTICS THINKING QUESTIONS

Exercise 15

Which city generates the highest total trip_fare?

Exercise 16

Which vehicle_type has the highest average fare?

Exercise 17

Which driver has completed the most trips?

Exercise 18

Why is Parquet preferred for analytics dashboards and aggregations?

OPTIONAL CHALLENGE

Challenge 1

Repartition the dataset into 4 partitions and write to Parquet.

Challenge 2

Create a summary dataset with:

- city
- total_trips
- total_revenue

- average_trip_duration

Write it to Parquet.
