```
#check that java is installed
!java -version
→ openjdk version "11.0.28" 2025-07-15
    OpenJDK Runtime Environment (build 11.0.28+6-post-Ubuntu-1ubuntu122.04.1)
    OpenJDK 64-Bit Server VM (build 11.0.28+6-post-Ubuntu-1ubuntu122.04.1, mixed mode, sharing)
#install pyspark
!pip install pyspark
    Requirement already satisfied: pyspark in /usr/local/lib/python3.11/dist-packages (3.5.1)
     Requirement already satisfied: py4j==0.10.9.7 in /usr/local/lib/python3.11/dist-packages (from pyspark) (0.10.9.7)
# -----
# STEP 1: Install & Import
# ==============
!pip install pyspark --quiet
from pyspark.sql import SparkSession
from pyspark.sql.functions import lit, rand, floor, concat_ws, explode, array, col, count
import time
# Create Spark Session
spark = SparkSession.builder \
   .appName("SaltingExample") \
   .config("spark.sql.shuffle.partitions", 8) \
   .getOrCreate()
# STEP 2: Create Skewed Sales Data
# -----
# Fact table: sales (skew on customer_id=1)
num_rows = 5_000_000
skew_ratio = 0.8 # 80% rows belong to customer_id=1
# Create skewed data
data_skewed = []
for i in range(num_rows):
   if i < num_rows * skew_ratio:</pre>
       data_skewed.append((1, i % 100, float(i % 500))) # customer_id=1
       cust\_id = i \% 1000 + 2 \# other customers
       data_skewed.append((cust_id, i % 100, float(i % 500)))
sales_df = spark.createDataFrame(data_skewed, ["customer_id", "product_id", "amount"])
# Dimension table: customers
customers = [(1, "VIP Customer")] + [(i, f"Customer_{i}") for i in range(2, 1002)]
customers_df = spark.createDataFrame(customers, ["customer_id", "customer_name"])
# -----
# STEP 3: Show Skew Distribution
# ===============
sales_df.groupBy("customer_id").agg(count("*").alias("cnt")) \
   .orderBy(col("cnt").desc()) \
   .show(5)
    Customer Distribution (Skewed):
    +------
     |customer_id| cnt|
              1 4000000
              2
                   1000
              12
              26
                   1000
              28
                   1000
    only showing top 5 rows
```

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                                                        Resolving_OOM_using_Salting.ipynb - Colab
    # -----
   # STEP 4: Skewed Join (No Salting)
   # -----
   start = time.time()
   joined_skewed = sales_df.join(customers_df, "customer_id")
   joined_skewed.count() # Force execution
   end = time.time()
   print(f" ∑ Join Time (Skewed, No Salting): {round(end - start, 2)} sec")
    🚁 🟅 Join Time (Skewed, No Salting): 10.88 sec
    # Partition size distribution before salting
   partition_sizes_skewed = joined_skewed.rdd.mapPartitions(lambda it: [sum(1 for _ in it)]).collect()
   print("\n Partition Sizes (Skewed):", partition_sizes_skewed)
    ∓
         Partition Sizes (Skewed): [2499584, 2500416]
    # -----
    # STEP 5: Salting the Join
    # -----
   salt size = 10 # Break heavy key into 10 parts
   # Add salt to customers table
    customers_salted = customers_df.withColumn("salt", floor(rand() * salt_size)) \
       .withColumn("join_key", concat_ws("_", col("customer_id"), col("salt")))
    # Add all salt values for sales table
    .withColumn("join_key", concat_ws("_", col("customer_id"), col("salt")))
   # Join on salted key
    start = time.time()
    joined_salted = sales_salted.join(customers_salted, "join_key") \
       .drop("join_key", "salt")
    joined_salted.count() # Force execution
   end = time.time()
   print(f" / Join Time (With Salting): {round(end - start, 2)} sec")
```

→ Join Time (With Salting): 47.13 sec

Double-click (or enter) to edit

```
# -----
# STEP 6: Cleanup
spark.stop()
Start coding or generate with AI.
```

https://colab.research.google.com/drive/1eUsjOZDnZQFHoUYQWXZn_RklrYqyHIYF#scrollTo=r2uk-Niqhlbr&printMode=true