



## 1. PySpark API

- ◆ PySpark provides an API that allows users to work with Apache Spark using Python. It enables handling big data efficiently with a distributed computing framework.

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```
from pyspark.sql import SparkSession

spark =
SparkSession.builder.appName("PySpark
Example").getOrCreate()

# Creating a DataFrame
data = [("Alice", 25), ("Bob", 30),
("Charlie", 35)]
df = spark.createDataFrame(data,
["Name", "Age"])

# Using SQL-like operations
df.filter(df.Age > 28).show()
```

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## ⚡ 2. Project Tungsten (Memory & CPU Optimization)

◆ Project Tungsten is Spark's effort to improve execution speed by optimizing memory, CPU efficiency, and code generation.

◆ Why Tungsten Matters?

- ✓ Up to 10x faster than older Spark versions
- ✓ Reduces garbage collection overhead
- ✓ Minimizes serialization & deserialization cost



```
spark.conf.set("spark.sql.execution.arrow.pyspark.enabled", "true")
df = spark.range(1,
1000000).toDF("num")
df.selectExpr("num * 2 as double_num").show()
```

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### 🔥 3. Catalyst Optimizer (Query Optimization Engine)

- ◆ Catalyst Optimizer is Spark's query optimization engine that improves execution plans for better performance.

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## ◆ Four Phases of Catalyst Optimizer

### 1. Analysis:-

Parses query, resolves column names, and verifies data types.

### 2. Logical Optimization:-

Rewrites query for efficiency (e.g., pushing filters down).

### 3. Physical Planning:-

Generates an optimized execution plan.

### 4. Code Generation (Tungsten Integration):-

Generates JVM bytecode for efficient execution.