

Actividad Módulo 47 - Big Data Parte 1

Generar un archivo .pdf que contenga las salidas:

- Configuración de plataforma Spark
- Importación de datos de Housing a una estructura de Spark
- Selección de datos de housing con filtros simples:
 1. Listado completo de columnas
 2. Para el zipcode con mayor número de casas, calcular el promedio de precio y tamaño en m2
- Agrupamiento en Spark, por zipcode, por número de habitaciones y baños, precio promedio

```
In [ ]: from pyspark.sql import SparkSession

# Create Spark Session
spark = SparkSession\
    .builder\
    .appName('ActividadMod47')\
    .getOrCreate()
```

```
In [ ]: # Importación de datos de Housing
df = spark.read.csv('D:/Documentos/Documentos/DataAnalysis/ebac/Python/Modulo44/kc_
df.show(10)
```

```

+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
|      id|      date|      price|bedrooms|bathrooms|sqft_living|sqft_lot|floors|waterfront|view|condition|grade|sqft_above|sqft_basement|yr_built|yr_renovated|zip|
|      ipcode|      lat|      long|sqft_living15|sqft_lot15|
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
|7129300520|20141013T000000|221900|3|1|1180|5650|
1|0|0|3|7|1180|0|1955|0|
98178|47.5112|-122.257|1340|5650|
|6414100192|20141209T000000|538000|3|2.25|2570|7242|
2|0|0|3|7|2170|400|1951|1991|
98125|47.721|-122.319|1690|7639|
|5631500400|20150225T000000|180000|2|1|770|10000|
1|0|0|3|6|770|0|1933|0|
98028|47.7379|-122.233|2720|8062|
|2487200875|20141209T000000|604000|4|3|1960|5000|
1|0|0|5|7|1050|910|1965|0|
98136|47.5208|-122.393|1360|5000|
|1954400510|20150218T000000|510000|3|2|1680|8080|
1|0|0|3|8|1680|0|1987|0|
98074|47.6168|-122.045|1800|7503|
|7237550310|20140512T000000|1.225e+006|4|4.5|5420|101930|
1|0|0|3|11|3890|1530|2001|0|
98053|47.6561|-122.005|4760|101930|
|1321400060|20140627T000000|257500|3|2.25|1715|6819|
2|0|0|3|7|1715|0|1995|0|
98003|47.3097|-122.327|2238|6819|
|2008000270|20150115T000000|291850|3|1.5|1060|9711|
1|0|0|3|7|1060|0|1963|0|
98198|47.4095|-122.315|1650|9711|
|2414600126|20150415T000000|229500|3|1|1780|7470|
1|0|0|3|7|1050|730|1960|0|
98146|47.5123|-122.337|1780|8113|
|3793500160|20150312T000000|323000|3|2.5|1890|6560|
2|0|0|3|7|1890|0|2003|0|
98038|47.3684|-122.031|2390|7570|
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+

```

only showing top 10 rows

```

In [ ]: # Selección de datos de housing
        # 1) Listado completo de columnas
        df.printSchema()

```

```
root
|-- id: string (nullable = true)
|-- date: string (nullable = true)
|-- price: string (nullable = true)
|-- bedrooms: string (nullable = true)
|-- bathrooms: string (nullable = true)
|-- sqft_living: string (nullable = true)
|-- sqft_lot: string (nullable = true)
|-- floors: string (nullable = true)
|-- waterfront: string (nullable = true)
|-- view: string (nullable = true)
|-- condition: string (nullable = true)
|-- grade: string (nullable = true)
|-- sqft_above: string (nullable = true)
|-- sqft_basement: string (nullable = true)
|-- yr_built: string (nullable = true)
|-- yr_renovated: string (nullable = true)
|-- zipcode: string (nullable = true)
|-- lat: string (nullable = true)
|-- long: string (nullable = true)
|-- sqft_living15: string (nullable = true)
|-- sqft_lot15: string (nullable = true)
```

```
In [ ]: # Para el zipcode con mayor número de casas, calcular el promedio de precio y tamaño
        from pyspark.sql.types import FloatType, IntegerType, BooleanType

        # Cast columns as the correct data type
        df = df.withColumn('price', df.price.cast(FloatType()))
        df = df.withColumn('bedrooms', df.bedrooms.cast(IntegerType()))
        df = df.withColumn('bathrooms', df.bathrooms.cast(FloatType()))

        df = df.withColumn('sqft_living', df.sqft_living.cast(IntegerType()))
        df = df.withColumn('sqft_lot', df.sqft_lot.cast(IntegerType()))
        df = df.withColumn('floors', df.floors.cast(IntegerType()))
        df = df.withColumn('waterfront', df.waterfront.cast(BooleanType()))
        df = df.withColumn('view', df.view.cast(BooleanType()))
        df = df.withColumn('condition', df.condition.cast(IntegerType()))
        df = df.withColumn('grade', df.grade.cast(IntegerType()))
        df = df.withColumn('sqft_above', df.sqft_above.cast(IntegerType()))
        df = df.withColumn('sqft_basement', df.sqft_basement.cast(IntegerType()))

        df = df.withColumn('lat', df.lat.cast(FloatType()))
        df = df.withColumn('long', df.long.cast(FloatType()))

        df = df.withColumn('sqft_living15', df.sqft_living15.cast(IntegerType()))
        df = df.withColumn('sqft_lot15', df.sqft_lot15.cast(IntegerType()))
```

```
In [ ]: # Para el zipcode con mayor número de casas, calcular el promedio de precio y tamaño
df.createOrReplaceTempView('KC_HOUSING')

sql_str = """
    select  zipcode,
            count(distinct id) as id_count,
            avg(price) as avg_price,
            avg(sqft_living) * 0.0929 as sqft_living_m2
    from KC_HOUSING
    group by zipcode
    order by count(distinct id) desc
    limit 3
"""
spark.sql(sql_str).show(10)
```

zipcode	id_count	avg_price	sqft_living_m2
98103	600	584919.2109634551	153.36215946843853
98038	587	366867.6	199.52274711864408
98115	576	619900.5471698113	170.498907890223

```
In [ ]: # Agrupamiento en Spark, por zipcode, por número de habitaciones y baños, precio pr
df.createOrReplaceTempView('KC_HOUSING')

sql_str = """
    select  zipcode,
            bedrooms,
            bathrooms,
            avg(price) as avg_price
    from KC_HOUSING
    group by zipcode, bedrooms, bathrooms
    order by 1,2,3
"""
spark.sql(sql_str).show(20)
```

zipcode	bedrooms	bathrooms	avg_price
98001	0	0.0	139950.0
98001	1	1.0	166000.0
98001	1	2.0	171000.0
98001	2	1.0	197428.57142857142
98001	2	1.5	350000.0
98001	2	1.75	246112.5
98001	2	2.5	214100.0
98001	2	2.75	239475.0
98001	3	0.75	363000.0
98001	3	1.0	205182.80952380953
98001	3	1.5	224108.5
98001	3	1.75	260531.0810810811
98001	3	2.0	256841.42857142858
98001	3	2.25	265999.0
98001	3	2.5	308581.8604651163
98001	3	2.75	255000.0
98001	3	3.0	309500.0
98001	4	1.0	229790.0
98001	4	1.5	246406.85714285713
98001	4	1.75	251114.2857142857

only showing top 20 rows