ENERGY DEMAND ANALYSIS IN SPAIN

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GOALS AND OBJECTIVES

MOTIVATION

- Forecasting in energy markets is one exceedingly helpful tool in making the transition to a renewable-based electrical infrastructure
- Our goal is to demonstrate this by leveraging Big Data analysis tools on a dataset that consists of energy usage and weather data for five large cities in Spain

SIGNIFICANCE

- Predict energy usage to increase efficiency of electrical production
- Predict energy price
- Locate areas that would benefit from renewable energies

OBJECTIVES

- Predict energy usage based on the weather
- Predict energy prices by:
 - Time of day
 - Day of the week
 - Time of year
- Analyze the factors that affect the fluctuations in energy usage, as well as the sources of energy

FEATURES

- dt_iso (datetime index localized to CET)
- generation biomass (in MW)
- generation fossil brown coal/lignite (in MW)
- generation fossil coal-derived gas (in MW)
- generation fossil gas (in MW)
- generation fossil hard coal (in MW)
- generation fossil oil (in MW)
- generation fossil oil shale (in MW)
- generation fossil peat (in MW)

- generation geothermal (in MW)
- city_name
- temp (in kelvin)
- temp_min (in kelvin)
- temp_max (in kelvin)
- pressure (in hPa)
- humidity (in %)
- wind_speed (in m/s)
- wind_deg (wind direction)
- rain_Ih (rain in last hour in mm)
- ect...

STORYTELLING - CHAPTER I

WHO?

• This data has the ability to impact every community that uses an electrical grid. Not only is it advantageous at the individual level to be able to predict the cost of an electric bill, but it is also extremely helpful to be able to predict energy usage at a macro level as communities across the globe begin to make the transition to renewable energies in response to climate change.

WHAT?

• We seek to analyze the patterns of weather and energy uses and use them to predict energy demand and price. While the short-term results have the potential to improve 24-hour and hour-by-hour predictions, this work also has the potential to help identify areas that would most benefit from the introduction of renewable energy.

WHEN? / WHERE?

- This data was collected between 2015 2019.
- The data is collected from the five largest cities in Spain: Madrid, Barcelona, Valencia, Seville, and Bilbao.

WHY?

• By improving forecasting, we can increase the efficiency of a power grid and help reduce the usage of peak demand on power plants, which are generally less efficient than their counterparts. As stated in the motivation, forecasting in energy markets is an exceedingly helpful tool in making the transition to a renewable-based electrical infrastructure.

CURRENT WORK

HIVE

- Load data into Hive using Cloudera
- Timestamp conversion
- Table merging
- Querying

city_name	temp	temp_min	temp_max
Barcelona	289.84594528884395	288.59273345304103	291.0194106526
ilbao	286.3761614905216	284.914581781618	288.0341095718221
Madrid	288.0581355640967	286.8221485281031	289.1525875339992
Seville	293.1028779257481	291.1815026537708	295.959789556759
/alencia	290.77789969745476	290.21955934550954	291.3519373417158

x.city_name	x.year	temp
Barcelona	2015	289.8164228233073
Barcelona	2016	289.9148484257806
Barcelona	2017	289.7989752190377
Barcelona	2018	289.85331927296517
Bilbao	2015	286.3070560449736
Bilbao	2016	286.51417122701736
Bilbao	2017	286.2761192779457
Bilbao	2018	286.40737739985383
Madrid	2015	288.14126680611423
Madrid	2016	288.0550533660679
Madrid	2017	288.6491765270033
Madrid	2018	287.413208712464
Seville	2015	293.9569442655581
Seville	2016	294.4828176623545
Seville	2017	293.36447039732354
Seville	2018	290.6301364922095
Valencia	2015	290.03898830741883
Valencia	2016	290.45791672346394
Valencia	2017	291.2427910806921
Valencia	2018	291.3730365067371
	+	++-

```
SELECT x.city_name, x.year,
AVG(x.temp) as temp
FROM (
SELECT city_name, YEAR(time)
AS year, temp
FROM joined
) X
GROUP BY x.city_name,
x.year;
```

.city_name	x.year	temp
Barcelona	2015	289.8164228233073
Barcelona	2016	289.9148484257806
Barcelona	2017	289.7989752190377
Barcelona	2018	289.85331927296517
ilbao	2015	286.3070560449736
ilbao	2016	286.51417122701736
ilbao	2017	286.2761192779457
ilbao	2018	286.40737739985383
adrid	2015	288.14126680611423
ladrid	2016	288.0550533660679
adrid	2017	288.6491765270033
adrid	2018	287.413208712464
eville	2015	293.9569442655581
eville	2016	294.4828176623545
Seville	2017	293.36447039732354
eville	2018	290.6301364922095
alencia	2015	290.03898830741883
alencia	2016	290.45791672346394
alencia	2017	291.2427910806921
alencia	2018	291.3730365067371

0: jdbc:hive2://>

LIMITATIONS

- Solr / Lucene great for indexing, searching full-text data in a high volume setting
- Our dataset contains numerical data rather than text
- Other tools fit our use case better

FUTURE WORK

FUTURE WORK

- MapReduce sophisticated query/analysis with custom MapReduce .jar
- Cassandra build table in Cassandra and run query
- Sqoop transfer data from HDFS to MySQL and run query

```
mysql> SHOW TABLES;
 Tables in group project
  energy
1 row in set (0.00 sec)
mysql> DESCRIBE energy;
 Field
                                            Type
                                                        | Null | Key | Default
            Extra
                                             | timestamp | NO
                                                                      CURRENT
 TIMESTAMP | on update CURRENT TIMESTAMP |
 generation biomass
                                              float
                                                         YES
                                                                      NULL
                                             float
 generation fossil brown coal lignite
                                                        YES |
                                                                      NULL
 generation fossil coal derived gas
                                              float
                                                        YES |
                                                                      NULL
                                             float
 generation fossil gas
                                                         YES |
                                                                      NULL
 generation fossil hard coal
                                             float
                                                        YES |
                                                                      NULL
                                             float
 generation fossil oil
                                                        YES
                                                                      NULL
                                             float
                                                        | YES |
                                                                      NULL
 generation fossil oil shale
 generation fossil peat
                                              float
                                                         YES
                                                                      NULL
 generation geothermal
                                             float
                                                        YES
                                                                      NULL
 generation hydro pumped storage aggregated
                                              float
                                                        YES
                                                                      NULL
 generation hydro pumped storage consumption | float
                                                        YES |
                                                                      NULL
                                              float
                                                         YES
 generation hydro run of river and poundage
                                                                      NULL
```