


05_Step5_Summary and comparisons of Bear and Weber WEAP Models

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Execute the following cells by pressing **Shift-Enter** , or by pressing the play button  on the toolbar above.

1. Import python libraries

```
In [1]: # 1. Import python libraries
#### set the notebook mode to embed the figures within the cell
import numpy
import sqlite3
import numpy as np
import pandas as pd
import getpass
from hs_restclient import HydroShare, HydroShareAuthBasic
import os

import plotly
plotly.__version__
import plotly.offline as offline
import plotly.graph_objs as go
from plotly.offline import download_plotlyjs, init_notebook_mode, plot, iplot
offline.init_notebook_mode(connected=True)
from plotly.offline import init_notebook_mode, iplot
from plotly.graph_objs import *

init_notebook_mode(connected=True) # initiate notebook for offline plot

import os
import csv
from collections import OrderedDict
import sqlite3
import pandas as pd
import numpy as np
from IPython.display import display, Image, SVG, Math, YouTubeVideo
import urllib
import calendar

print 'The needed Python libraries have been imported'
```

The needed Python libraries have been imported

2. Connect to the WaMDaM populated SQLite file

```
In [2]: # 2. Connect to the WaMDaM populated SQLite file

# Then we can run queries against it within this notebook :)

# the SQLite file is published here
#https://github.com/WamdhamProject/WaMDaM_UseCases/blob/master/UseCases_files/3SQLite_database/BearRiverDataset
WaMDaM_SQLite_Name='Bear_Weber.sqlite'

# WaMDaM_SQLite_Name='replicateWaMDaM.sqlite'

conn = sqlite3.connect(WaMDaM_SQLite_Name)

print 'Connected to the WaMDaM SQLite file called'+': '+ WaMDaM_SQLite_Name

Connected to the WaMDaM SQLite file called: Bear_Weber.sqlite
```

Summary and comparisons of Bear and Weber WEAP Models

```
In [3]: # Use Case 3.1Identify_aggregate_TimeSeriesValues.csv
# plot aggregated to monthly and converted to acre-feet time series data of multiple sources

# 2.2Identify_aggregate_TimeSeriesValues.csv
Query_UseCase3_1_URL="""
https://raw.githubusercontent.com/WamdhamProject/WaMDaM_JupyterNotebooks/master/2_ServeToModels/SQL/GetModelSummary
"""

# Read the query text inside the URL
Query_UseCase3_1_text = urllib.urlopen(Query_UseCase3_1_URL).read()

# return query result in a pandas data frame
result_df_UseCase3_1= pd.read_sql_query(Query_UseCase3_1_text, conn)

# uncomment the below line to see the list of attributes
display (result_df_UseCase3_1)

print "Queries are done"
```

	MasterNetworkName	ScenarioName	AttributeDataTypeCV	CountOfAttributes	CountOfInstances
0	Bear River Network	Bear River WEAP Model 2017	FreeText	89	61
1	Bear River Network	Bear River WEAP Model 2017	MultiAttributeSeries	11	11
2	Bear River Network	Bear River WEAP Model 2017	NumericValues	281	197
3	Bear River Network	Bear River WEAP Model 2017	SeasonalNumericValues	36	33
4	Bear River Network	Bear River WEAP Model 2017	TimeSeries	37	36
5	Weber	Base	FreeText	32	30
6	Weber	Base	MultiAttributeSeries	8	8
7	Weber	Base	NumericValues	464	152
8	Weber	Base	SeasonalNumericValues	15	15
9	Weber	Base	TimeSeries	44	44

Queries are done

Total demand

```
In [4]: # Use Case 3.1Identify_aggregate_TimeSeriesValues.csv
# plot aggregated to monthly and converted to acre-feet time series data of multiple sources

# 2.2Identify_aggregate_TimeSeriesValues.csv
Query_UseCase3_1_URL="""
https://raw.githubusercontent.com/WamdhamProject/WaMDaM_JupyterNotebooks/master/2_ServeToModels/SQL/GetAverageAnnualDemand
"""

# Read the query text inside the URL
Query_UseCase3_1_text = urllib.urlopen(Query_UseCase3_1_URL).read()

# return query result in a pandas data frame
result_df_UseCase3_1= pd.read_sql_query(Query_UseCase3_1_text, conn)

# uncomment the below line to see the list of attributes
display (result_df_UseCase3_1)

print "Queries are done"
```

	ScenarioName	AttributeDataTypeCV	NoumInstances	AverageAnnualDemand
0	Base	TimeSeries	19	473385.212121
1	Bear River WEAP Model 2017	NumericValues	3	777799.090000
2	Bear River WEAP Model 2017	SeasonalNumericValues	21	990552.939225

Queries are done

Total Discharge

```
In [5]: # Use Case 3.1Identify_aggregate_TimeSeriesValues.csv
# plot aggregated to monthly and converted to acre-feet time series data of multiple sources

# 2.2Identify_aggregate_TimeSeriesValues.csv
Query_UseCase3_1_URL="""
https://raw.githubusercontent.com/WamdhamProject/WaMDaM_JupyterNotebooks/master/2_ServeToModels/SQL/GetTotalHeadflow
"""

# Read the query text inside the URL
Query_UseCase3_1_text = urllib.urlopen(Query_UseCase3_1_URL).read()

# return query result in a pandas data frame
result_df_UseCase3_1= pd.read_sql_query(Query_UseCase3_1_text, conn)

# uncomment the below line to see the list of attributes
display (result_df_UseCase3_1)

print "Queries are done"
```

	ResourceTypeAcronym	ScenarioName	#Instance	AttributeDataTypeCV	AttributeName_Abstract	AnnualDischarge
0	WEAP_Bear	Bear River WEAP Model 2017	11	TimeSeries	Surface Water Inflow	1.265191e+06
1	WEAP_Bear	Bear River WEAP Model 2017	12	TimeSeries	Headflow	1.026013e+06
2	WEAP_Bear	Bear River WEAP Model 2017	Dewitt Spring Headflow	SeasonalNumericValues	Headflow	1.059976e+04
3	WEAP_Weber	Base	17	TimeSeries	Headflow	9.423588e+05
4	WEAP_Weber	Base	Q68 Sewage Outfall Headflow	SeasonalNumericValues	Headflow	3.314313e+04

Queries are done

1 Cubic Feet Per Second to Acre-feet Per Month = 60.3707

Total Capacity

```
In [6]: # Use Case 3.1Identify_aggregate_TimeSeriesValues.csv
# plot aggregated to monthly and converted to acre-feet time series data of multiple sources

# 2.2Identify_aggregate_TimeSeriesValues.csv
Query_UseCase3_1_URL="""
https://raw.githubusercontent.com/WamdhamProject/WaMDaM_JupyterNotebooks/master/2_ServeToModels/SQL/GetTotalReservoirCapacity
"""

# Read the query text inside the URL
Query_UseCase3_1_text = urllib.urlopen(Query_UseCase3_1_URL).read()

# return query result in a pandas data frame
result_df_UseCase3_1= pd.read_sql_query(Query_UseCase3_1_text, conn)

# uncomment the below line to see the list of attributes
display (result_df_UseCase3_1)

print "Queries are done"
```

	ScenarioName	#Reservoirs	TotalCapacity
0	Base	8	551240.000
1	Bear River WEAP Model 2017	10	140411.002
2	Bear River WEAP Model 2017	Bear Lake	1516633.000

Queries are done

```
In [ ]:
```