```
In [1]: # Import libraries
from pandas import DataFrame, concat, date_range, rolling_mean
from sqlalchemy import create_engine, MetaData, Table, select
import matplotlib.pyplot as plt
```

Grab Data from SQL

In this section we use the *sqlalchemy* library to grab data from ADEPT. The table we are concerned about is located in the *Marketing* database and called *IEICustomerCounts*. Cindea/Karen are the team members that maintain this table and BI is just mainly using the data in the table. This table is the source of the data behind the customer count numbers Brad in sales uses.

```
In [2]: # Parameters
         ServerName = "RepSer2"
        Database = "Marketing"
         TableName = "IEICustomerCounts"
         # Create the connection
         engine = create engine('mssql+pyodbc://' + ServerName + '/' + Database)
         conn = engine.connect()
         # Required for querying tables
        metadata = MetaData(conn)
         # Table to query
         tbl = Table (TableName, metadata, autoload=True, schema="dbo")
         #tbl.create(checkfirst=True)
         # Select all
         sql = tbl.select()
         # run sql code
         result = conn.execute(sql)
         # Insert to a dataframe
         df = DataFrame(data=list(result), columns=result.keys())
         # Convert data types
        df.StatusDate = df.StatusDate.astype('datetime64[ns]')
        df.CustomerCount = df.CustomerCount.astype('int')
        print ' '
        print 'Data Types'
        print df.dtypes
         # Set index to dataframe
         df = df.set index('StatusDate', drop=False)
         # Close connection
         conn.close()
```

object

Data Types State

```
StatusDate
                   datetime64[ns]
T<sub>1</sub>DC
                            object
Pool
                            object
IsCGS
                            object
IsUCB
                            object
CustomerType
                            object
Status
                            object
                             int64
CustomerCount
```

```
In [3]: df.head()
```

Out[3]:

	State	StatusDate	LDC	Pool	IsCGS	IsUCB	CustomerType	Status	CustomerCount
StatusDate									
2011-05-16	GA	2011-05-16 00:00:00	AGLC	SAV	Non-CGS	N/A	Residential	Inactive	896
2011-05-16	GA	2011-05-16 00:00:00	AGLC	SNG	Non-CGS	N/A	Residential	Inactive	1694
2011-05-16	GA	2011-05-16 00:00:00	AGLC	TRA	Non-CGS	N/A	Residential	Inactive	938
2011-05-16	GA	2011-05-16 00:00:00	AGLC	VAL	Non-CGS	N/A	Residential	Inactive	94
2011-05-16	GA	2011-05-16 00:00:00	AGLC	None	CGS	N/A	Residential	Pending	177

Do some Data Munging

This section attempts to clean up the data for analysis.

- 1. Only select records where the account status is equal to "Active"
- 2. Make sure the state column is all in upper case
- 3. Merge (NJ and NY) to NY in the state column
- 4. Remove any ouliers (I found some zero and other odd results in the data set)

```
In [4]: # Only grab where Status == Active
    df = df[df['Status'] == 'Active']

# Clean State Column, convert to upper case
    df.State = df.State.apply(lambda x: x.upper())

# Convert NJ to NY
    df.State[df.State == 'NJ'] = 'NY'

# Group by State and StatusDate
    Daily = df.groupby(['State', 'StatusDate']).sum()

# Calculate Outliers
    StateYearMonth = Daily.groupby([Daily.index.get_level_values(0), Daily.index.get_level_values(1).year, Daily.index.get_level_values(1).month])
    Daily['Outlier'] = StateYearMonth['CustomerCount'].transform( lambda x: abs(x-x.mean()) > 1.60*x.std())

# Remove Ouliers
    Daily = Daily[Daily['Outlier'] == False]
```

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The dataframe named **Daily** will hold customer counts that have been aggregated per day. The original data (tbl IEICustomerCounts) has multiple records per day. We are left with a data set that is indexed by both the state and the StatusDate. The Oulier column should be equal to zero signifying that the record is not an outlier.

```
In [5]: Daily.head()
```

Out[5]:

		CustomerCount	Outlier
State	StatusDate		
	2009-09-05	9948	0
	2009-09-12	9937	0
FL	2009-09-19	9919	0
	2009-09-26	9916	0
	2009-10-03	9924	0

We create a seperate dataframe named *ALL* which groups the Daily dataframe by StatusDate. We are essentially getting rid of the State column. The *Max* column represents the maximum customer count per month. The Max column is used to smooth out the graph.

```
In [6]: # Combine all markets

# Get the max customer count by Date
ALL = DataFrame (Daily['CustomerCount'].groupby (Daily.index.get_level_values(1)).sum())
ALL.columns = ['CustomerCount'] # rename column

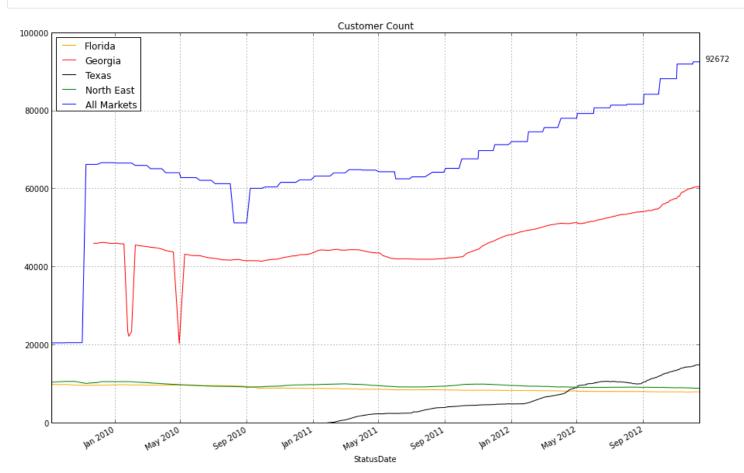
# Group by Year and Month
YearMonth = ALL.groupby([lambda x: x.year, lambda x: x.month])

# What is the max customer count per Year and Month
ALL['Max'] = YearMonth['CustomerCount'].transform(lambda x: x.max())
ALL.head()
```

Out[6]:

	CustomerCount	Max
StatusDate		
2009-09-05	20501	20626
2009-09-12	20543	20626
2009-09-19	20560	20626
2009-09-26	20626	20626
2009-10-03	20671	20671

Graph Customer Counts per Market



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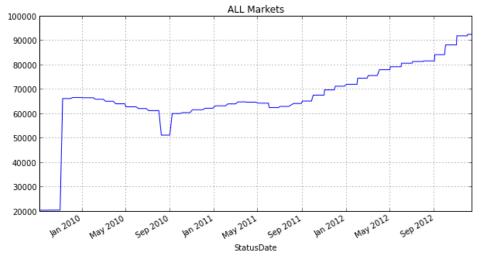
Graph Customer Counts, create sub plots per Market

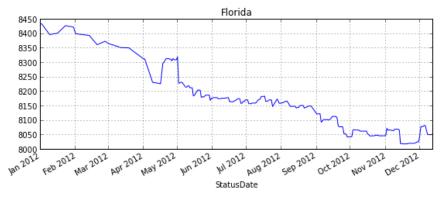
```
In [8]: # First Graph
ALL['Max'].plot(figsize=(10, 5));plt.title('ALL Markets')

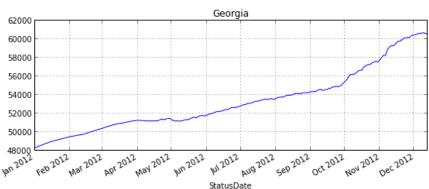
# Last four Graphs
fig, axes = plt.subplots(nrows=2, ncols=2, figsize=(20, 10))
fig.subplots_adjust(hspace=1.0) ## Create space between plots

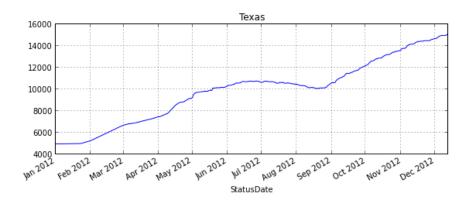
Daily.ix['FL']['CustomerCount']['2012':].fillna(method='pad').plot(ax=axes[0,0]); axes[0,0].set_title('Florida')
Daily.ix['GA']['CustomerCount']['2012':].fillna(method='pad').plot(ax=axes[0,1]); axes[0,1].set_title('Georgia')
Daily.ix['TX']['CustomerCount']['2012':].fillna(method='pad').plot(ax=axes[1,0]); axes[1,0].set_title('Texas')
Daily.ix['NY']['CustomerCount']['2012':].fillna(method='pad').plot(ax=axes[1,1]); axes[1,1].set_title('North East')
```

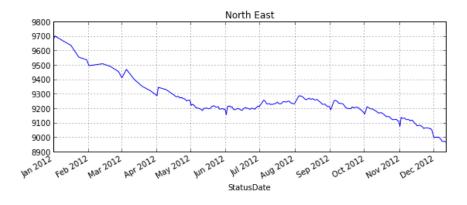
Out[8]: <matplotlib.text.Text at 0x92ecfd0>











Add the BHAG numbers to the "ALL" dataframe

12/31/2012 - 92,000 customers 12/31/2013 - 138,600 customers 12/31/2014 - 200,000 customers

```
In [9]: # Create the BHAG dataframe
    data = [92000,138600,200000]
    idx = date_range(start='12/31/2012', end='12/31/2014', freq='A')
    BHAG = DataFrame(data, index=idx, columns=['BHAG'])
    BHAG
```

Out[9]:

	BHAG
2012-12-31	92000
2013-12-31	138600
2014-12-31	200000

```
In [10]: # Combine the BHAG and ALL data sets
  combined = concat([ALL,BHAG], axis=0)
  combined.tail()
```

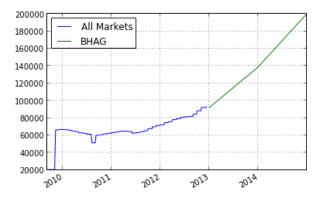
Out[10]:

	BHAG	CustomerCount	Max
2012-12-12	NaN	92672	92672
2012-12-13	NaN	83783	92672
2012-12-31	92000	NaN	NaN
2013-12-31	138600	NaN	NaN

```
2014-12-31 | 200000 | NaN | NaN
```

```
In [11]: combined['Max'].plot(color='blue', label='All Markets')
    combined['BHAG'].plot(color='green', label='BHAG')
    plt.legend(loc='best')
```

Out[11]: <matplotlib.legend.Legend at 0x938ac50>



Do some basic Forecasting

```
In [12]: # Group by Year and then get the max value per year
Year = combined.groupby(lambda x: x.year).max()

# Add a column representing the percent change per year
Year['YR_PCT_Change'] = Year['Max'].pct_change(periods=1)
Year
```

Out[12]:

	BHAG	CustomerCount	Max	YR_PCT_Change
2009	NaN	66794	66794	NaN
2010	NaN	66705	66705	-0.001332
2011	NaN	71426	71426	0.070774
2012	92000	92672	92672	0.297455
2013	138600	NaN	NaN	NaN
2014	200000	NaN	NaN	NaN

To get next year's end customer count we will assume our current growth rate remains constant. We then will increase this years customer count by that amount and that will be our forecast for next year.

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```
In [13]: (1 + Year.ix[2012,'YR_PCT_Change']) * Year.ix[2012,'Max']
```

Out[13]: 120237.72273401843