Final Proposal: The Effects of Subsidence on Houston, Texas

Principal Investigator: Ava Ring

Email: aring@nyu.edu

This project will examine the compaction of land in the Houston, Texas area. A recent hurricane, Hurricane Harvey, devestated the area and highlighted the fact that most of Houston is below sea level and prone to flooding. This peaked my interest and led me to ask, is Houston sinking?

There are many reasons parts of the earth's surface will sink, but the most prevelant is *subsidence*. "Land subsidence is a gradual settling or sudden sinking of the Earth's surface owing to subsurface movement of earth materials." In most cases, subsidence is caused from the extraction of water for drinking or agricultural purposes from layers of rock. When the water is extracted, it causes the layers to smash together, compact, and then sink. Subsidence occurs most frequently in dry, arid places where extracting water from aquifers is a necessity. Besides Texas, Nevada and California also contain large subsidence bowls More information on subsidence can be found https://water.usgs.gov/ogw/subsidence.html).

Let me give you an illustrative example of subsidence. In the 1950s, a group of oil tycoons moved to an area south east of Houston city center, which they titled "Oil Bay." With the new residents came increased development of the land and increased extraction of water from a nearby aquifer. Because Houston is already below sea-level and so flat, when a major hurricane hit (Alicia) the whole town was under water. It had sunk over 30 feet in a few decades.

Houston is the largest subsidence bowl in the United States, and the first to be studied. After Oil Bay, Harris-County created Harris-Galveston Subsidence District (HGSD) in 1975. "The District was created to provide for the regulation of groundwater withdrawal throughout Harris and Galveston counties for the purpose of preventing land subsidence, which leads to increased <u>flooding</u> (https://hgsubsidence.org/about-the-district/)."

This project will examine the positive effects Harris-Galveston Subsidence District (HGDS) has had on subsidence and land compaction in Houston. It will show how water extraction from aquifers leads to land subsindence. It will demonstrate how subsidence has decreased since HGDS's founding and regulations have taken place.

Vocababulary to know:

- aquifer a body of permeable rock that can contain or transmit groundwater.
- borehole extensometer measurement tool used to determine the movement behavior of soil and rock masses

Data Report

Land compaction was recorded using 13 extensometers near the Chicot and Evangeline aquifers in the Houston area. The extensometer sites were establed from 1973 to 1980. Measurements of compaction are recorded about 13 times a year until 2015.

The data can be found at <u>USGS.gov (https://pubs.usgs.gov/sim/3365/tables/)</u>.

```
In [60]: import pandas as pd
   import matplotlib.pyplot as plt
   import numpy as np
   import matplotlib.dates as mdates
   from mpl_toolkits.axes_gridl.inset_locator import zoomed_inset_axes

%matplotlib inline
```

- pandas is used to make the dataframes of extensometer data.
- matplotlib.pyplot is used to plot the data.
- · numpy is used for manipulation of the data.
- matplotlib.dates is for putting datetime series data on the graph's x axis.
- mpl_toolkits.axes_grid1.inset_locator is needed for a graph inset.

Step One: Reading in the Data

Out[61]: 'sim3365_table4D.xlsx'

```
In [62]: # combining the path to the website and the specific tables' urls to make a
         urls = [(usgs + item) for item in data_list]
         urls
Out[62]: ['https://pubs.usgs.gov/sim/3365/tables/sim3365 table4A.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4B.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365_table4C.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4D.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4E.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365_table4F.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4G.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4H.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4I.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4J.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4K.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4L.xlsx',
          'https://pubs.usgs.gov/sim/3365/tables/sim3365 table4M.xlsx']
```

Now we are reading in the excel sheets.

Out[63]:

	Measurement date	Compaction (feet)
0	1980-07-22	0.000
1	1980-08-19	0.008
2	1980-09-16	0.014
3	1980-10-14	0.002
4	1980-11-10	0.025

In [64]: # we still need to fix the column names so they are easier to work with.
lukily, all the dataframes are structured the same way, so a for loop fixe
I also set the index to the Measurement date since this is a time series of

for df in list_of_dfs:
 df.columns = ["Measurement_date", "Compaction"]

 df.set_index("Measurement_date", inplace = True)

list_of_dfs[6].head()

Out[64]:

Compaction

Measurement_date	
1973-07-13 00:00:00	0.000
1973-08-13 00:00:00	0.007
1973-09-13 00:00:00	0.005
1973-10-11 00:00:00	0.010
1973-11-08 00:00:00	0.012

```
In [65]: # now I name the 13 dataframes according to their location.
         # I set a string name for them as well so that I can easily identify each or
         lake_houston = list_of_dfs[0] #tableA
         lake_houston.name = "lake_houston"
         northeast = list_of_dfs[1] #tableB
         northeast.name = "northeast"
         southwest = list_of_dfs[2] #tableC
         southwest.name = "southwest"
         east_end = list_of_dfs[3] #tableD
         east_end.name = "east_end"
         addicks = list_of_dfs[4] #tableE
         addicks.name = "addicks"
         johnson_space = list_of_dfs[5] #tableF
         johnson space.name = "johnson space"
         texas_city = list_of_dfs[6] #tableG
         texas_city.name = "texas_city"
         baytown_shallow = list_of_dfs[7] #tableH
         baytown_shallow.name = "baytown_shallow"
         baytown deep = list of dfs[8] #tableI
         baytown deep.name = "baytown deep"
         seabrook = list_of_dfs[9] #tableJ
         seabrook.name = "seabrook"
         clear_lake_shallow = list_of_dfs[10] #tableK
         clear_lake_shallow.name = "clear_lake_shallow"
         clear lake deep = list of dfs[11] #tableL
         clear_lake_deep.name = "clear_lake_deep"
         pasadena = list of dfs[12] #tableM
         pasadena.name = "pasadena"
```

Let's take a look at them all.

	Compaction
Measurement_date	
1976-05-26	0.000
1976-06-09	0.004
1976-06-21	0.002
1976-07-20	0.009
1976-08-16	0.021

pasadena

	Compaction
Measurement_date	
1975-10-08	0.000
1975-10-20	0.010
1975-11-20	0.026
1975-12-22	0.040
1976-01-21	0.054

Problem with Data

The dataframes look pretty good now, with a few exceptions.

Texas City

This dataframe has a different index that consists of strings and not DateTime Index. I have to look further into this because if I want to use a function on it, the index has to be in DateTimeIndex form.

Different Start Dates

For all of the dataframes (besides Johnson Space), the data starts at 0 when the first measurement of subsidence was taken. Then, it measures the compaction in feet relative to that start point.

The problem is that the first measurement was taken at different years for some of the sites. For example, Lake Houston subsidence measures start in 1980, while Addicks data starts in 1973. So if I compare Lake Houston subsidence vs Addicks subsidense in 2015, it will seem like Addicks has sunk significantly more compared to the other sites. However, it will only seem this way because Addicks data includes 7 additional years of measurements.

The dataframes i have to change are southwest, east_end, addicks, johnson_space, texas_city, baytown_shallow, baytown_deep, seabrook, clear_lake_shallow, clear_lake_deep and pasadena.

Johnson Space

Johnson Space dataframe is also a bit strange because it does not start with a measurement of 0, but of 1.347. I will rework this dataframe so it starts with 0 and goes from there, so it is consistent with the other sites.

Step 2: Fix Texas City

```
In [67]: texas_city.head()
```

Out[67]:

Compaction

Measurement_date	
1973-07-13 00:00:00	0.000
1973-08-13 00:00:00	0.007
1973-09-13 00:00:00	0.005
1973-10-11 00:00:00	0.010
1973-11-08 00:00:00	0.012

Texas City has the time, which is a red flag. The index type is below.

```
In [68]: print(type(texas_city.index))
    print(type(addicks.index))

# this shows that the index is not right.
# I need a datetime index for texas_city, like for example addicks.

<class 'pandas.core.indexes.base.Index'>
```

<class 'pandas.core.indexes.datetimes.DatetimeIndex'>

so i read in the data again so that I can fix the problem from the beginning.

```
In [69]: texasurl = 'https://pubs.usgs.gov/sim/3365/tables/sim3365_table4G.xlsx'
    new_texas_city = pd.read_excel(texasurl,skiprows = [0, 1, 2], skip_footer =
```

In [70]: new_texas_city.tail(10)

Out[70]:

	Measurement date	Compaction (feet)
542	2015-03-30 00:00:00	0.097
543	2015-04-29 00:00:00	0.096
544	2015-05-19 00:00:00	0.097
545	2015-06-22 00:00:00	0.097
546	7/23/201	0.099
547	2015-09-08 00:00:00	0.100
548	2015-09-29 00:00:00	0.096
549	2015-10-28 00:00:00	0.097
550	2015-11-23 00:00:00	0.096
551	2015-12-21 00:00:00	0.095

Here you can see row 546 was a typo. Therefore, we have to change that entry to a date. In the data, "7/21/201" instead of "7/21/2015." This is why the column "Measurement date" was considered a string object, and not a datetime series.

```
In [71]: new_texas_city.columns = ["Measurement_date", "Compaction"]

# I change the name of the columns so they are easier to work with

new_texas_city.loc[546, "Measurement_date"] = "2015-07-23 00:00:00"

# then I change row 546 to the correct date

new_texas_city.tail(10)
```

Out[71]:

	Measurement_date	Compaction
542	2015-03-30 00:00:00	0.097
543	2015-04-29 00:00:00	0.096
544	2015-05-19 00:00:00	0.097
545	2015-06-22 00:00:00	0.097
546	2015-07-23 00:00:00	0.099
547	2015-09-08 00:00:00	0.100
548	2015-09-29 00:00:00	0.096
549	2015-10-28 00:00:00	0.097
550	2015-11-23 00:00:00	0.096
551	2015-12-21 00:00:00	0.095

5/13/2018

```
new_texas_city["Measurement_date"] = pd.to_datetime(new_texas_city["Measurement_date"])
In [73]: # Then I set the index to this column and set it inplace True.

new_texas_city.set_index("Measurement_date", inplace = True)

texas_city = new_texas_city

print(texas_city.head())

print(type(texas_city.index))
```

In [72]: # now I can convert the column "Measurement date" to a DateTime form, instead

Step 3: Fix Different Start Dates

First, I truncate the dataframes to start at July 1980.

```
In [74]: # I create a function I can apply on each of the dataframes.
# This function tells it to drop any entries with DateTimeIndexes before Jul

def dropyears(df_):
    return(df_.truncate(before = "1980-7"))
```

```
In [75]: southwest = southwest.apply(dropyears)
    east_end = east_end.apply(dropyears)
    addicks = addicks.apply(dropyears)
    johnson_space = johnson_space.apply(dropyears)
    texas_city = texas_city.apply(dropyears)
    baytown_shallow = baytown_shallow.apply(dropyears)
    baytown_deep = baytown_deep.apply(dropyears)
    seabrook = seabrook.apply(dropyears)
    clear_lake_shallow = clear_lake_shallow.apply(dropyears)
    clear_lake_deep = clear_lake_deep.apply(dropyears)
    pasadena = pasadena.apply(dropyears)
```

In [76]: # test to double check the function worked correctly.
addicks.head()

Out[76]:

Compaction

Measurement_date	
1980-07-23	0.672
1980-08-19	0.690
1980-09-16	0.714
1980-10-14	0.735
1980-11-10	0.745

Next, I create a "diff_" column that takes the difference in subsidence between each measurement day.

```
5/13/2018
```

```
In [78]: # Im using the diff method to find the difference between each row of Compact
for df in list_to_change:
    df["diff_"] = df.Compaction.diff()

    df.fillna(0, inplace = True)

addicks.head()
```

Out[78]:

Compaction diff_

Measurement_date			
1980-07-23	0.672	0.000	
1980-08-19	0.690	0.018	
1980-09-16	0.714	0.024	
1980-10-14	0.735	0.021	
1980-11-10	0.745	0.010	

But I need to take the sum of the previous row of diff_ column and the current diff column. This operation will give me the actual Compaction for each time period.

```
In [79]: # I use the "cumsum" method on the "diff_" column to achieve my want.

for df in list_to_change:
    df["cumsum_"] = df.diff_.cumsum()

addicks.head() # to see the cumsum_ column.
```

Out[79]:

Compaction diff_ cumsum_

weasurement_date			
1980-07-23	0.672	0.000	0.000
1980-08-19	0.690	0.018	0.018
1980-09-16	0.714	0.024	0.042
1980-10-14	0.735	0.021	0.063
1980-11-10	0.745	0.010	0.073

Messurement date

```
In [80]: # drop the uncessary columns from the dataframes

for df in list_to_change:
    df.drop("Compaction", axis = 1, inplace = True)

    df.drop("diff_", axis = 1, inplace = True)

addicks.head() # to see the dropped columns
```

Out[80]:

cumsum_

Measurement_date	
1980-07-23	0.000
1980-08-19	0.018
1980-09-16	0.042
1980-10-14	0.063
1980-11-10	0.073

```
In [81]: # rename cumsum_ to "Compaction"

for df in list_to_change:
    df.columns = ["Compaction"]

addicks.head() # to see the final product
```

Out[81]:

Compaction

1980-07-23	0.000
1980-08-19	0.018
1980-09-16	0.042
1980-10-14	0.063
1980-11-10	0.073

Measurement date

Measurement_date

```
Compaction
```

```
In [83]: len(addicks.Compaction)
```

Out[83]: 456

To look at the final compaction of every site, I print the last row of each dataframe.

Texas City and Johnson Space land has elevated in the past 40 years. The rest of the locations have sunk anywhere from .6 to 3 inches.

In [84]: for df in alldfs_list:
 print(df.tail(1))

F((-//
	Compaction
Measurement date	
2015-12-28	0.608
	Compaction
Measurement date	L
2015-12-23	0.905
2013-12-23	Compaction
Measurement_date	Compaction
2015-12-23	1.627
2015-12-23	
	Compaction
Measurement_date	
2015-12-23	0.514
	Compaction
Measurement_date	
2015-12-23	2.994
	Compaction
Measurement_date	
2015-12-21	-1.215
	Compaction
Measurement_date	
2015-12-21	-0.087
2013 12 21	Compaction
Measurement_date	Compaction
2015-12-28	0.508
2015-12-28	
	Compaction
Measurement_date	
2015-12-28	0.263
	Compaction
Measurement_date	
2015-12-21	0.421
	Compaction
Measurement date	
2015-12-21	0.445
	Compaction
Measurement_date	-
2015-12-21	0.465
	Compaction
Measurement date	Compaction
2015-12-21	0.163
2013-12-21	0.103

Step 4: Plotting the Data

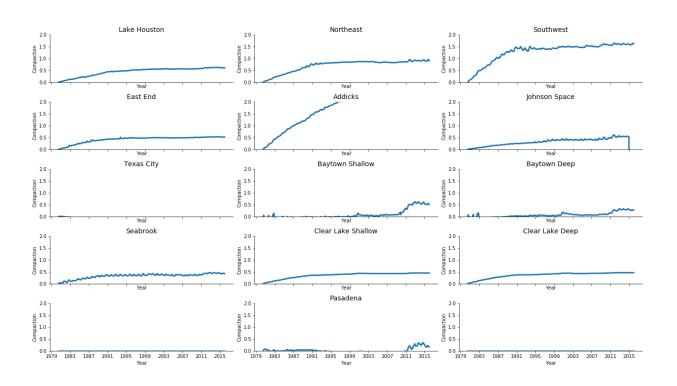
I want to see all of the trends of compaction at the different extensometer sites throughout Houston. Below I will plot all 13 sites.

I had to get a little creative here. An uneven amont of dataframes made sub-plotting difficult. For a work-around, I made up a fake dataframe with the same index as the other dataframes. I use the fake datafram (nonedf) as a filler for subplots 153, and 353.

```
nonedf = pasadena.copy() # this is me making a placeholder dataframe
        nonedf["Compaction "] = np.nan # I fill it with NaN values
        nonedf.Compaction .fillna(0, inplace = True) # then fill the NaN with 0s
        nonedf.drop("Compaction", axis = 1, inplace = True) # drop the initial column
        nonedf.columns = ["Compaction"] # and rename the NaN column to Compaction
        fig, ax = plt.subplots(nrows = 5, ncols = 3, sharex = True, figsize = (18,10
        ax = ax.ravel() # this lets me use a for loop on all of the dfs.
       var list = [lake houston, northeast, southwest, east end, addicks,
                  johnson_space, texas_city, baytown_shallow, baytown_deep,
                  seabrook, clear_lake_shallow, clear_lake_deep, nonedf, pasadena,
        # above are my dataframes.
        nice_name = ["Lake Houston", "Northeast", "Southwest", "East End", "Addicks"
                   "Johnson Space", "Texas City", "Baytown Shallow", "Baytown Deer
                   "Seabrook", "Clear Lake Shallow", "Clear Lake Deep", " ", "Pasa
        # this list I will use to name all of the charts.
        count = 0
        for xxx in ax:
           xxx.plot(var_list[count].Compaction, linewidth = 3.0)
           xxx.set title(nice name[count], fontsize = 14)
           xxx.spines["right"].set_visible(False) # remove the top spines
           xxx.spines["top"].set visible(False) # remove the right spines
           xxx.set xlabel("Year", fontsize = 10) # set the x label to Year
           xxx.set_ylabel("Compaction", fontsize = 10) # set the Y label to Compact
           xxx.set ylim(0.0, 2.0) # set the y axis limits
           count = count + 1 # update the count
        fig.suptitle("Subsidence", fontsize = 22, fontweight = "bold", y = 1.15)
```

above formats the title and adds an extra spacing between the title and gi
plt.tight_layout(pad=0.4, w_pad=0.5, h_pad=1.0)
above makes the graphs fit nicely.

Subsidence



Step 5: Interprutting the Data

This data provided by USGS highlights many interesting points about subsidence and regulation.

Effects of Water Extraction on Subsidence

• Seabrook supports the fact that water extraction leads to subsidence.

Effect of HGSD regulation

- HGSD regulation takes effect in 1990. Since then less water is being extracted from aquifers and therefore less subsidence occurs.
- The Addicks site is a perfect example of how well regulation of water extraction diminishes sinking since HGSD regulation takes place much later than the other sites.

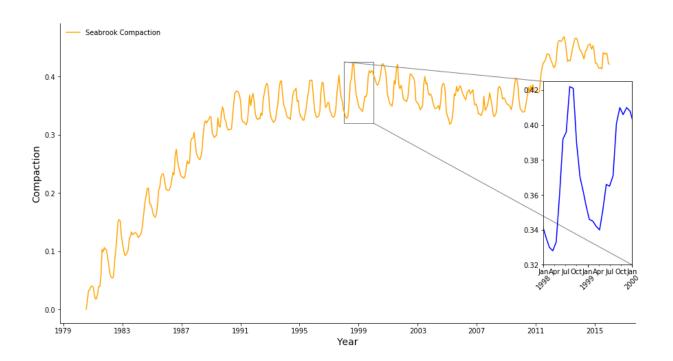
I will use additional graphs to shine further light on these points.

Seabrook

```
In [86]: fig, ax = plt.subplots(figsize = (15,8)) # creating the ax
        ax.plot(seabrook.Compaction, color = "orange") # plotting seabrook data
        axins = zoomed_inset_axes(ax, 3.0, loc=7)
        # above creates an axis on top of the original ax,
        # tells it how much to zoom in (3.0)
        # and tells it what position to be in (center right)
        axins.plot(seabrook["Compaction"], color = "b")
        # above plots the seabrook data inside the inset
        x1, x2, y1, y2 = "1998", "2000", 0.32, .425
        axins.set_xlim(x1, x2) # apply the x-limits
        axins.set ylim(y1, y2) # apply the y - limits
        # this sets the limits for what to show in the inset
        from mpl_toolkits.axes_grid1.inset_locator import mark_inset
        mark inset(ax, axins, loc1=2, loc2=4, fc="none", ec="0.5")
        ax.set title("Seabrook", fontsize = 14, fontweight = "bold", y = 1.15)
        ax.set ylabel("Compaction", fontsize = 14,)
        ax.set xlabel("Year", fontsize = 14,)
        # adding the title, and x and y labels
        ax.spines["top"].set visible(False)
        ax.spines["right"].set_visible(False)
        # making the spines invisible
        ax.legend(["Seabrook Compaction"],frameon=False)
        # creating the legend
        years = mdates.YearLocator() # grabbing the Year
        months = mdates.MonthLocator(range(1, 13),
                                  bymonthday=1, interval=3)
        # grabbing every three months
        yearsFmt = mdates.DateFormatter('%Y') # formatting the Year
        mthsFmt = mdates.DateFormatter('%b') # formatting the month
        axins.xaxis.set major locator(years)
        axins.xaxis.set minor locator(months)
        axins.xaxis.set major formatter(yearsFmt)
        axins.xaxis.set minor formatter(mthsFmt)
        plt.setp(axins.xaxis.get_majorticklabels(),
                rotation=45, y = -.02)
```

5/13/2018

Seabrook



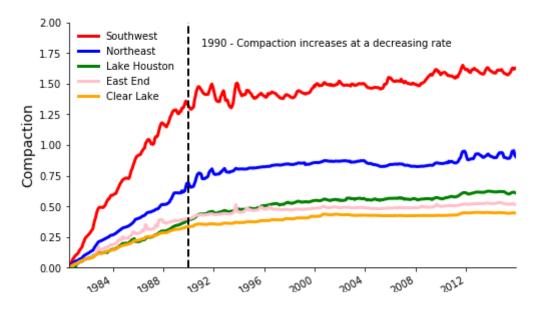
The above graph demonstrates seasonal subsidence. In the hot and dry months of the year (May to September) more water is extracted from the aquifer near this site. This leads to drier earth and more compaction. Whenever it is wetter and less water is being taken, the land rebounds and elevates. This pattern can be seen every year.

However, since HGDS regulation has gone into place, compaction has continued at a decreasing rate.

Decreasing rate of compaction

```
In [87]: fig, ax = plt.subplots(figsize = (8,5))
         southwest.Compaction.plot(ax = ax, linewidth = 3.0, color = "r")
         northeast.Compaction.plot(ax = ax, linewidth = 3.0, color = "b")
         lake_houston.Compaction.plot(ax = ax, linewidth = 3.0, color = "g")
         east_end.Compaction.plot(ax = ax, linewidth = 3.0, color = "pink")
         clear_lake_shallow.Compaction.plot(ax = ax, linewidth = 3.0, color = "orange
         ax.set_title("Decreasing Rate of Compaction", fontsize = 14, fontweight = "k
         ax.set_ylabel("Compaction", fontsize = 14,)
         ax.set_xlabel("Year", fontsize = 14,)
         ax.legend(["Southwest", "Northeast", "Lake Houston", "East End", "Clear Lake
                   frameon=False)
         ax.spines["top"].set_visible(False)
         ax.spines["right"].set_visible(False)
         ax.set_ylim(0.00, 2.00)
         # below I make the line for 1990
         ax.axvline(x="1990",
                    color='k',
                    linestyle='--',
                    linewidth=2)
         message = "1990 - Compaction increases at a decreasing rate"
         ax.text("1991", 1.8, message, horizontalalignment='left', fontweight = 1.5)
         plt.show()
```

Decreasing Rate of Compaction



The data on most sites show that the slopes of the graphs, the rate of compaction, was increasing at a faster rate in the 1980s and has shallowed out since the early 1990s. While subsidence has increased, it is not increasing at a decreasing rate. This can mostly be attributed to the founding of

Harris-Galveston Subsidence District (HGSD) in 1975.

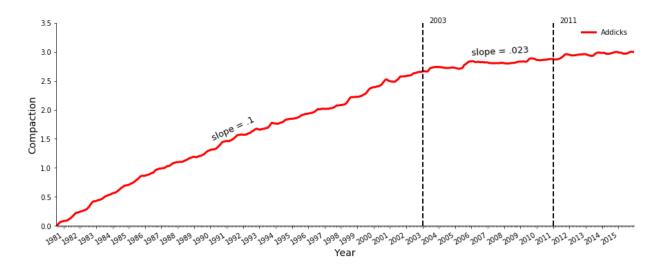
HGSD regulated the amount of water that could be extracted from the aquifers in the Harris-Galveston area. We can see a direct effect on when those regulations took place and the rates of compaction progressively decreasing.

A depiction of this phenomenon is provided above.

Addicks

```
In [88]: fig, ax = plt.subplots(figsize = (15,6))
         addicks.Compaction.plot(ax = ax,
                                linewidth = 3.0, color = "r")
         ax.set_title("Addicks", fontsize = 14, fontweight = "bold", y = 1.15)
         ax.set ylabel("Compaction", fontsize = 14,)
         ax.set_xlabel("Year", fontsize = 14,)
         ax.legend(["Addicks"], frameon=False)
         ax.spines["top"].set visible(False)
         ax.spines["right"].set_visible(False)
         ax.set_ylim(0.00, 3.50)
         years = mdates.YearLocator() # every year
         months = mdates.MonthLocator() # every month
         yearsFmt = mdates.DateFormatter('%Y')
         ax.xaxis.set_major_locator(years)
         ax.xaxis.set_major_formatter(yearsFmt)
         ax.xaxis.set_minor_locator(months)
         ax.axvline(x="2003",
                   color='k',
                    linestyle='--',
                    linewidth=2)
         ax.axvline(x= "2011",
                   color='k',
                    linestyle='--',
                    linewidth=2)
         message2003 = "2003"
         message2011 = "2011"
         ax.text("2003-6", 3.5, message2003, horizontalalignment='left', fontweight =
         ax.text("2011-6", 3.5, message2011, horizontalalignment='left', fontweight =
         first slope = "slope = .1"
         plt.text("1990", 1.8, first slope, fontsize = 13, rotation = 25)
         second slope = "slope = .023"
         plt.text("2006", 3, second slope, fontsize = 13, rotation = 3)
         plt.show()
```

Addicks



To further prove the positive effect of HGDS regulation, take a look at Addicks.

The addicks extensometer happens to be located in a area where HGDS regulation did not apply until 2011. Therefore, Addicks generally experienced higher levels of compaction than any other site from 1980 forward.

We can clearly see that compaction remained steady from 1980 to 2003, with about .1 feet of the earth compacting each measurement. In 2003, the USGS personnel deemed the adjacent well-supply where water was being pulled from as being inoperative. This curtailed the sinking of the land area. In 2011, the regulation went into effect.

Summary

Houston struggles from land subsidence because of it's arid and dry weather as well as its palcement below sea level. However, deep land water extraction from aquifers exasterbates this problem. This can be seen in the seasonal water extraction at Seabrook. HGDS's regulation has helped curtail the compaction of Houston since it's inception in 1975.

One long term solution to this problem is for residents and devlopments to get their water from surface supplies, instead of the ground. Natural resources like surrounding lakes and rivers should be utilized. However, this option is a more costly way of getting water.

Unless we want another Oil Bay on our hands, Houstonians should be aware of the effects of subsidence on the land they live on and the organizations helping to protect them.

In []:
