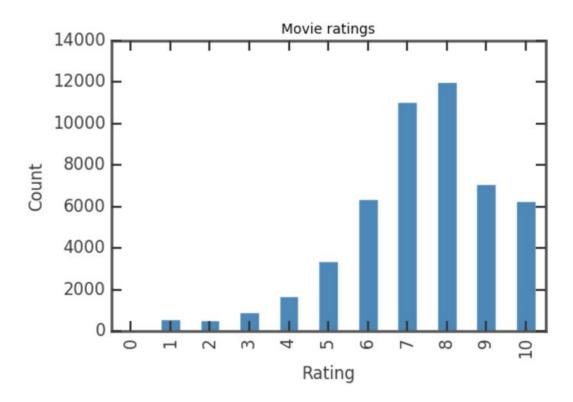
Chapter 1: Tools of the Trade

	user id	item id	rating	timestamp
0	1	1074638	7	1365029107
1	1	1853728	8	1366576639
2	2	104257	8	1364690142
3	2	1259521	8	1364118447
4	2	1991245	7	1364117717

```
12012
8
7
      11063
9
       7119
6
       6373
10
       6281
5
       3399
4
       1696
3
        924
1
        595
2
        533
0
           5
```

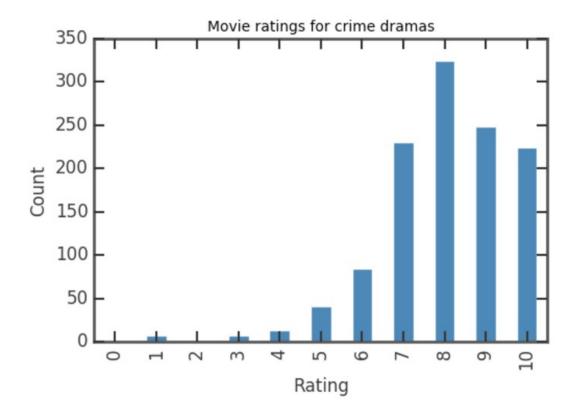
Name: rating, dtype: int64



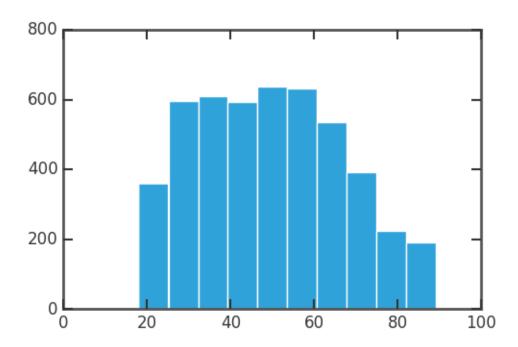
```
0 True
```

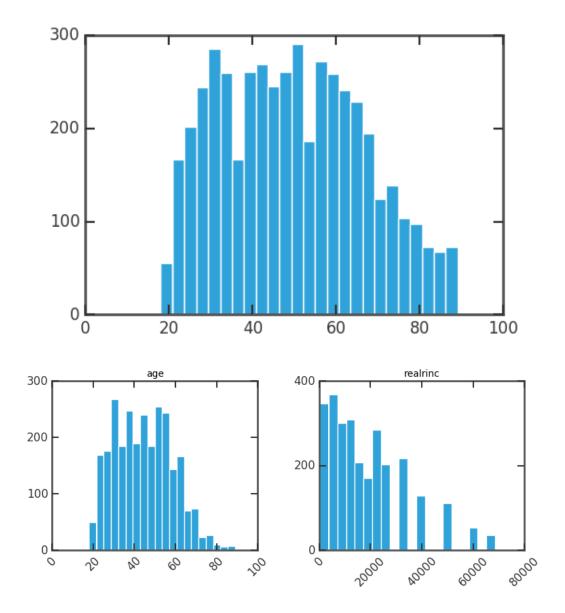
- 1 False
- 2 False
- 3 False
- 4 False

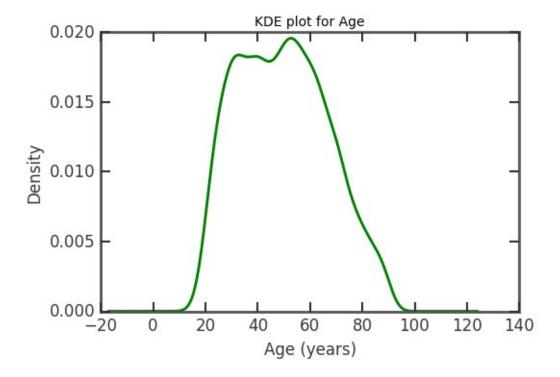
Name: genre, dtype: bool

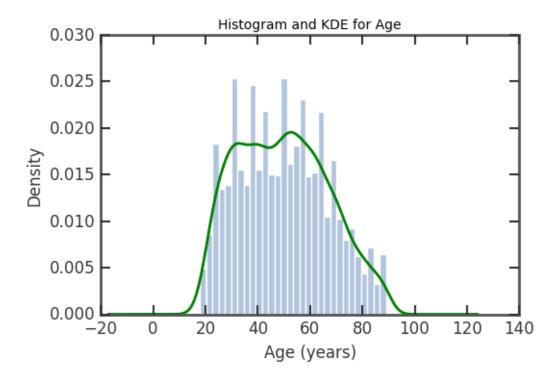


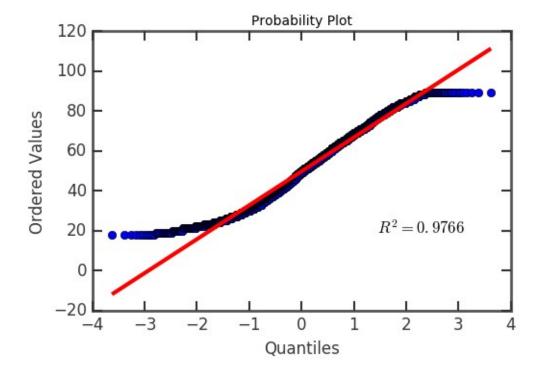
Chapter 2: Exploring Data

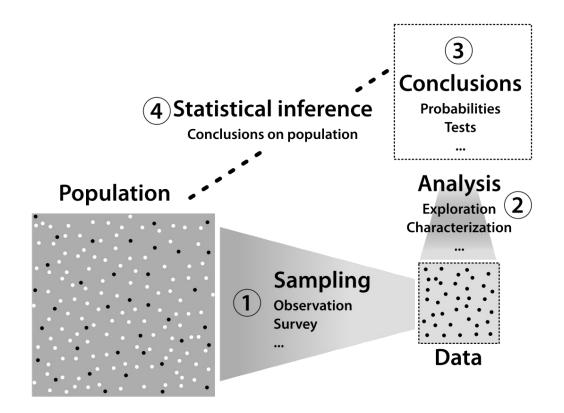


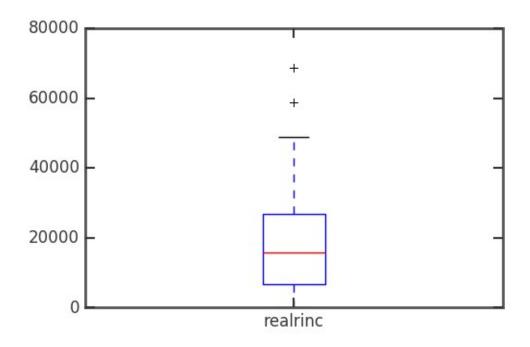


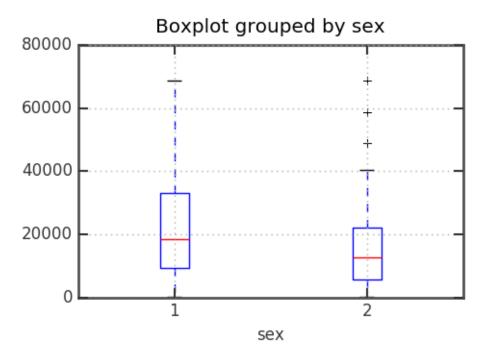


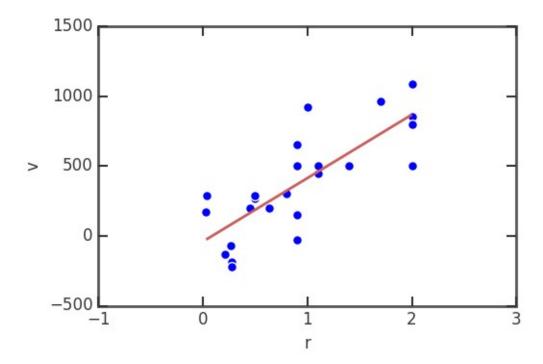




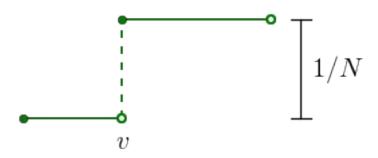


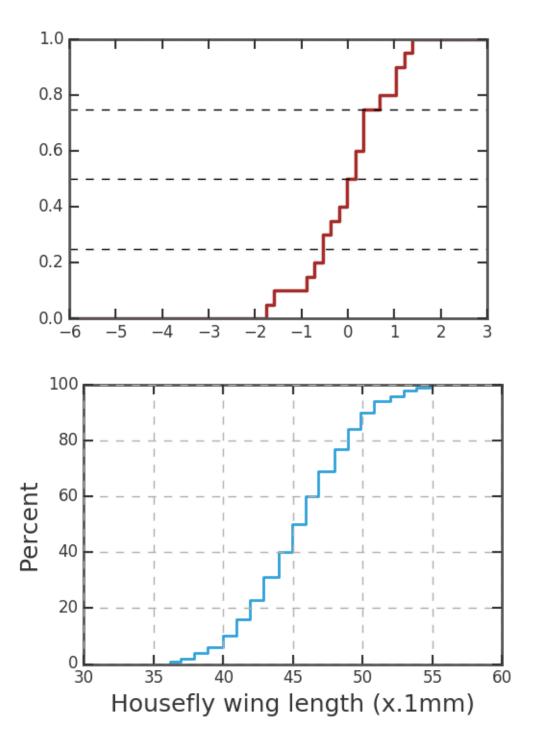


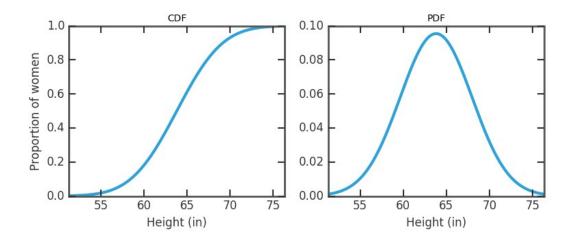


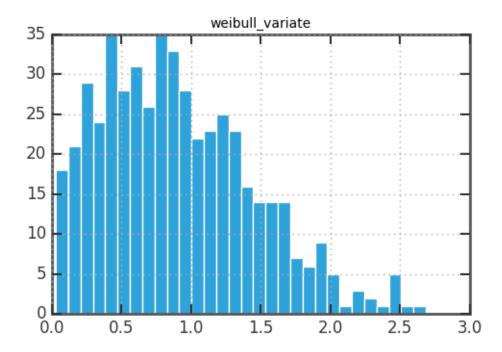


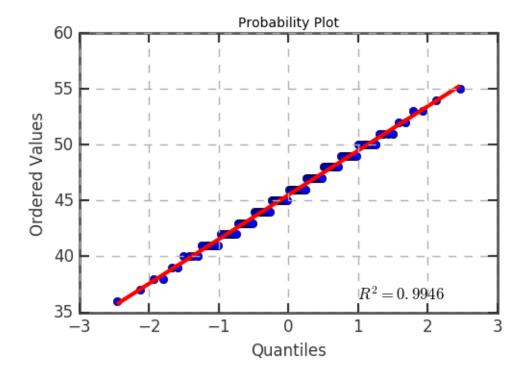
Chapter 3: Learning About Models

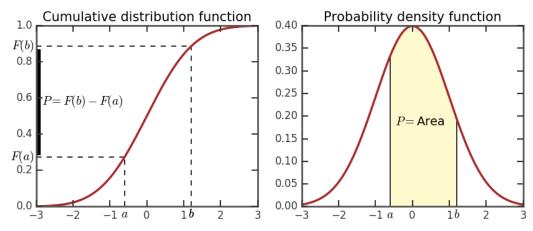


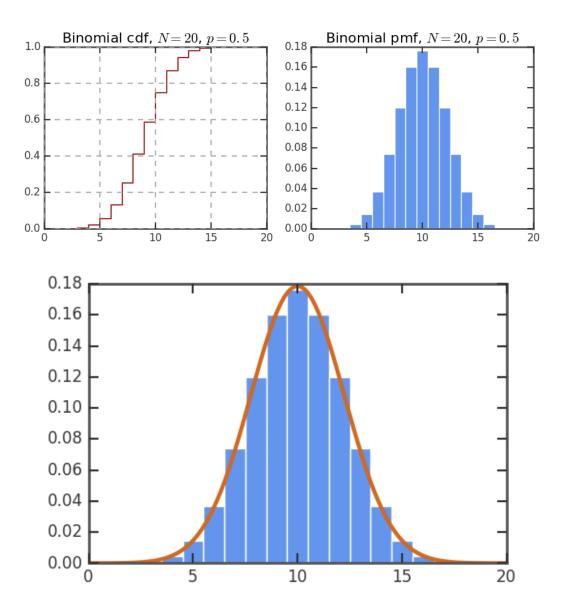


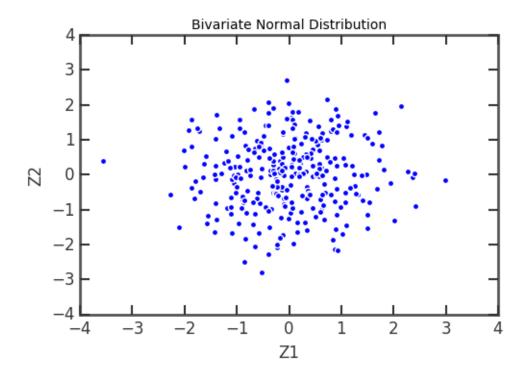






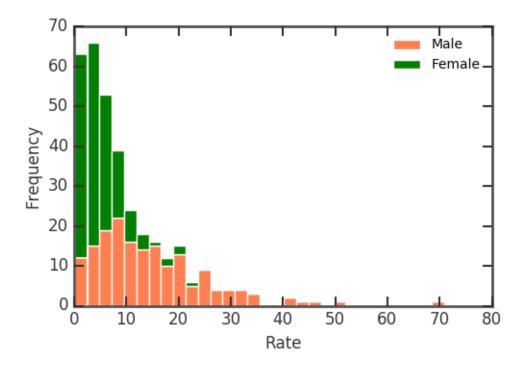


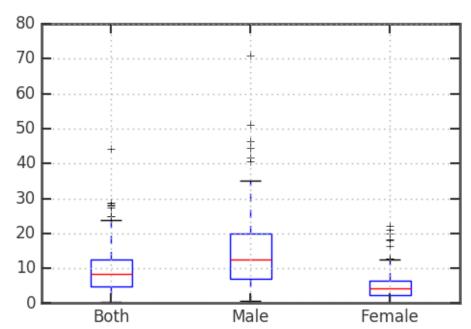


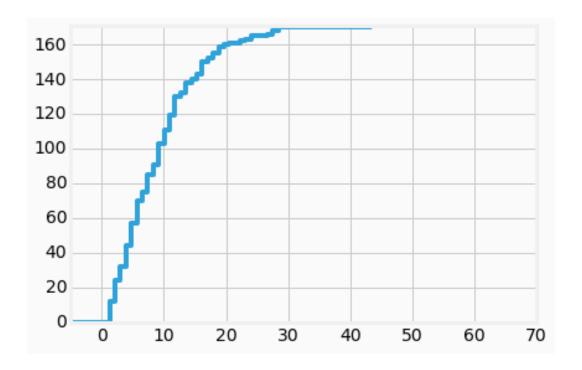


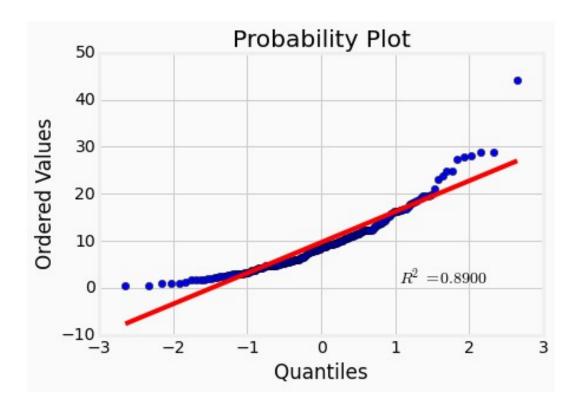
Chapter 4: Regression

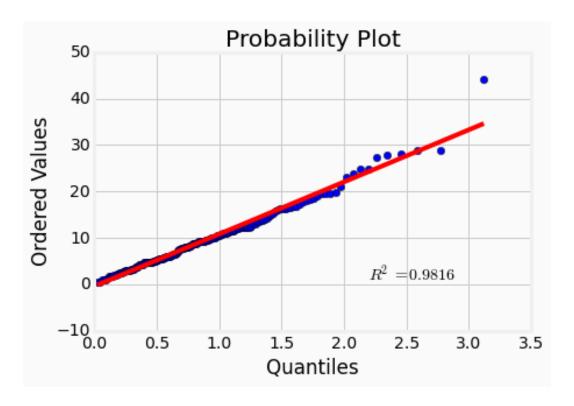
	Country	Both	Female	Male
0	Afghanistan	5.7	5.3	6.2
1	Albania	5.9	5.2	6.6
2	Algeria	1.9	1.5	2.3
3	Angola	13.8	7.3	20.7
4	Argentina	10.3	4.1	17.2
5	Armenia	2.9	0.9	5.0
6	Australia	10.6	5.2	16.1
7	Austria	11.5	5.4	18.2
8	Azerbaijan	1.7	1.0	2.4
9	Bahamas	2.3	1.3	3.6

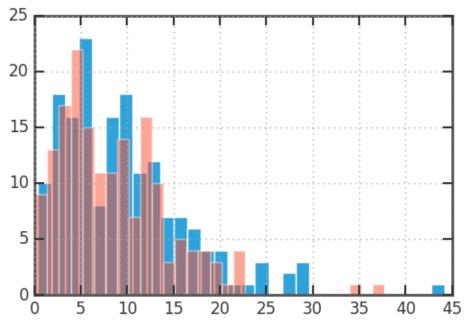


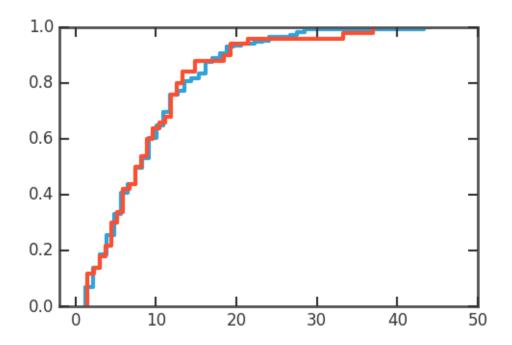






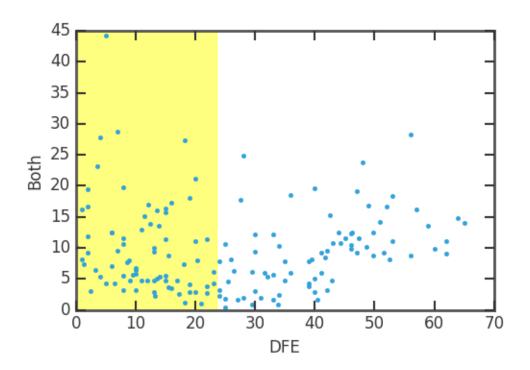


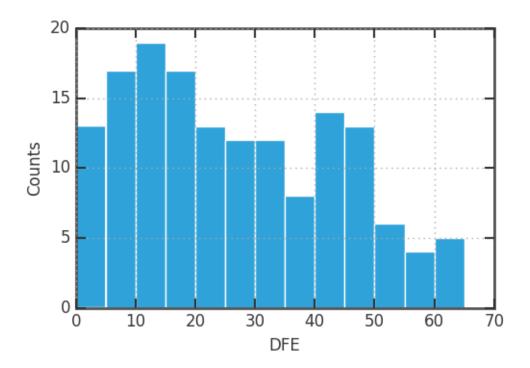


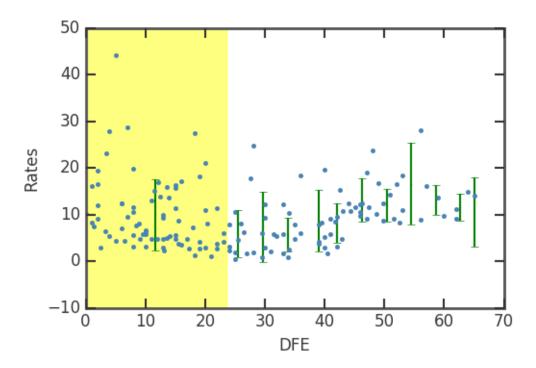


	LAT	LONG	DMS_LAT	DMS_LONG	MGRS	JOG	DSG	AFFIL	FIPS10	SHORT_NAME	FULL_NAME	MOD_DATE	ISO3136
0	33.000000	66.0	330000	660000	42STB1970055286	NI42- 09	PCLI	NaN	AF	Afghanistan	Islamic Republic of Afghanistan	2009-04-10	AF
1	41.000000	20.0	410000	200000	34TDL1589839239	NK34- 08	PCLI	NaN	AL	Albania	Republic of Albania	2007-02-28	AL
2	28.000000	3.0	280000	30000	31REL0000097202	NH31- 15	PCLI	NaN	AG	Algeria	People's Democratic Republic of Algeria	2011-03-03	DZ
3	-14.333333	-170.0	-142000	-1700000	1802701	NaN	PCLD	US	AS	American Samoa	Territory of American Samoa	1998-10-06	AS
4	42.500000	1.5	423000	13000	31TCH7675006383	NK31- 04	PCLI	NaN	AN	Andorra	Principality of Andorra	2007-02-28	AD

	Country	Both	Female	Male	Lat	Lon
0	Afghanistan	5.7	5.3	6.2	33	66
1	Albania	5.9	5.2	6.6	41	20
2	Algeria	1.9	1.5	2.3	28	3
3	Angola	13.8	7.3	20.7	-12.5	18.5
4	Argentina	10.3	4.1	17.2	-34	-64





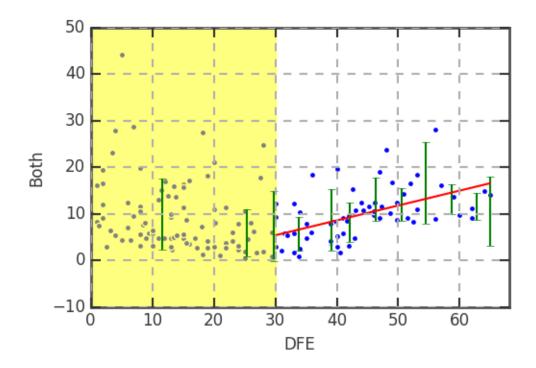


slope:0.3204 intercept:-4.2373 rvalue:0.5102 pvalue:0.0000 stderr:0.0715

OLS Regression Results

Dep. Variabl	e:		Both	R-sq	uared:		0.260	
Model:		OLS		Adj.	R-squared:		0.247	
Method:		Leas	t Squares	F-st	atistic:		20.06	
Date:		Mon, 21	Dec 2015	Prob	(F-statistic):	3.	.65e-05	
Time:			03:40:19	Log-	Likelihood:		-175.72	
No. Observat	ions:		59	AIC:			355.4	
Df Residuals	:		57	BIC:			359.6	
Df Model:			1					
Covariance T	ype:		nonrobust					
	coe	f std	err	t	P> t	[95.0% Conf	. Int.]	
Intercept	-4.237	3 3	. 272	-1.295	0.201	-10.789	2.315	
DFE	0.320	4 0	.072	4.479	0.000	0.177	0.464	
Omnibus:	======	======	13.615	Durb	======== in-Watson:		2.424	
Prob(Omnibus):		0.001	Jarq	ue-Bera (JB):		14.566	
Skew:	,		1.099		(JB): `´	0.	.000687	
Kurtosis:			4.047	Cond	. No.		238.	

Warnings: [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.



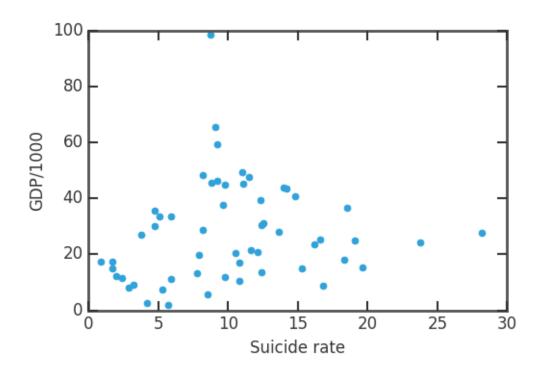
	id	name
680	6.0.GDPpc_constant	GDP per capita, PPP (constant 2011 internation
4611	GDPPCKD	GDP per Capita, constant US\$, millions
4612	GDPPCKN	Real GDP per Capita (real local currency units
6390	NE.GDI.FTOT.CR	GDP expenditure on gross fixed capital formati
6478	NV.AGR.PCAP.KD.ZG	Real agricultural GDP per capita growth rate (%)
6600	NY.GDP.PCAP.CD	GDP per capita (current US\$)
6601	NY.GDP.PCAP.CN	GDP per capita (current LCU)
6602	NY.GDP.PCAP.KD	GDP per capita (constant 2005 US\$)
6603	NY.GDP.PCAP.KD.ZG	GDP per capita growth (annual %)
6604	NY.GDP.PCAP.KN	GDP per capita (constant LCU)

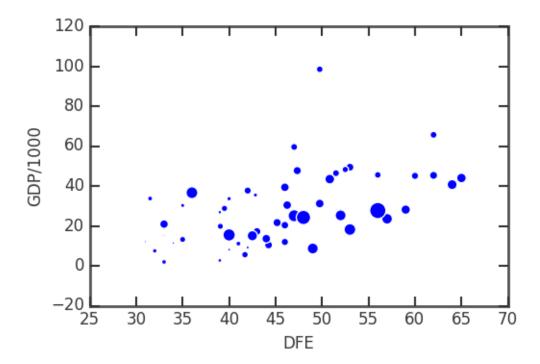
		NY.GDP.PCAP.PP.CD
country	year	
Arab World	2014	15975.039211
Caribbean small states	2014	15231.111124
Central Europe and the Baltics	2014	23884.797208
East Asia & Pacific (all income levels)	2014	14853.204148
East Asia & Pacific (developing only)	2014	11922.720831

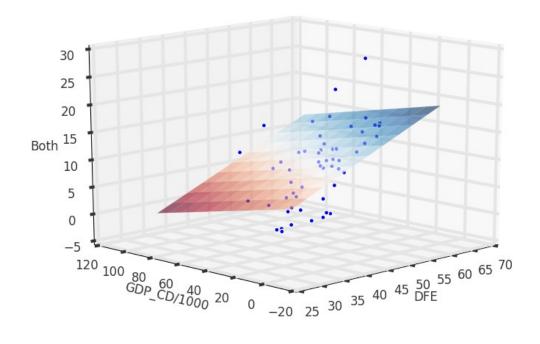
country	year	
Arab World	2014	15975.039211
Caribbean small states	2014	15231.111124
Central Europe and the Baltics	2014	23884.797208
East Asia & Pacific (all income le	vels) 2014	14853.204148
East Asia & Pacific (developing on	ly) 2014	11922.720831
Name: NY.GDP.PCAP.PP.CD, dtype: fl	oat64	
	country	gdp
0 An	ab World 1597	5.039210528601
1 Caribbean smal	1 states 1523	1.111124481498
2 Central Europe and the	Baltics 2388	4.797208032996
3 East Asia & Pacific (all income	levels) 1485	3.204148331899
4 East Asia & Pacific (developi	ng only) 11	922.7208309251

Country Both Female Male Lat Lon DFE GDP_CD 146 Sweden 11.1 6.1 16.2 62 15 62 45183 country gdp 218 Sweden 45183.0196872713 [45183.01968727]

	Country	Both	Female	Male	Lat	Lon	DFE	GDP_CD
146	Sweden	11.1	6.1	16.2	62	15	62	45183



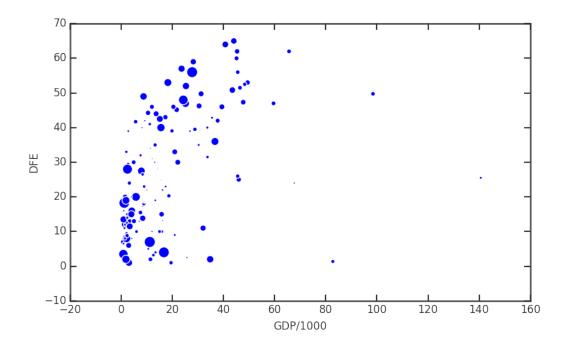




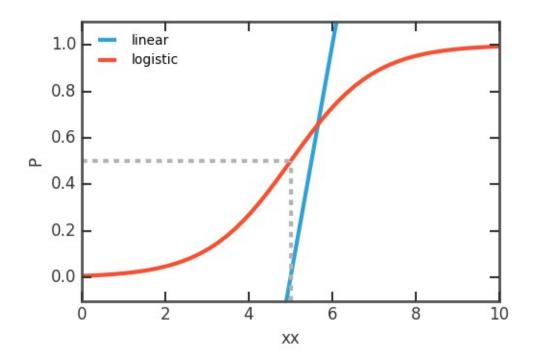
OLS Regression Results

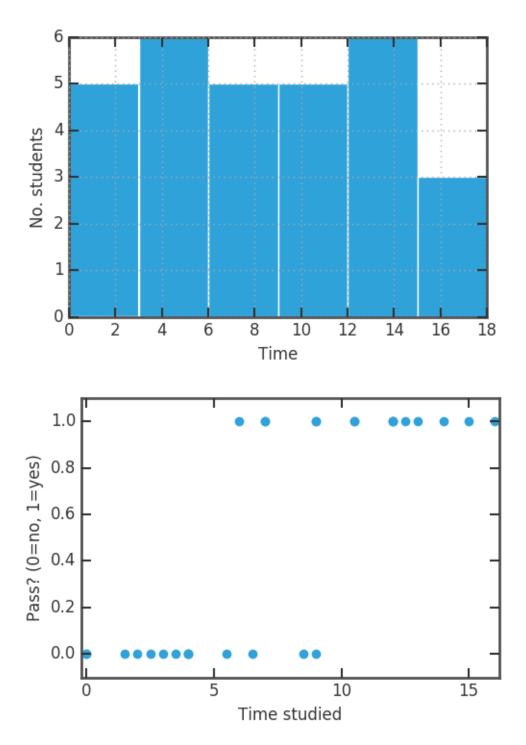
Dep. Vari	iable:	Во	th R-squa	red:		0.288
Model:		0	LS Adj. R	-squared:		0.260
Method:		Least Squar	es F-stat	istic:		10.33
Date:	1	Mon, 21 Dec 20	15 Prob (F-statistic):		0.000171
Time:		03:40:	42 Log-Li	kelihood:		-161.90
No. Obser	rvations:		54 AIC:			329.8
Df Residu	uals:		51 BIC:			335.8
Df Model:			2			
Covariano	ce Type:	nonrobu	st			
=======						
	coef		t	P> t	[95.0% Cor	nf. Int.]
const	-5.9298			0.108	-13.210	1.350
DFE	0.3942	0.090	4.397	0.000	0.214	0.574
GDP_CD	-6.238e-05	4.55e-05	-1.371	0.176	-0.000	2.9e-05
Omnibus:		11.8	36 Durbin	-Watson:	=======	2.272
Prob(Omni	ibus):	0.0	03 Jarque	-Bera (JB):		11.958
Skew:	•	1.0	62 Prob(J	B):		0.00253
Kurtosis:	:	3.8	99 Cond.	No.		1.72e+05

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified. [2] The condition number is large, 1.72e+05. This might indicate that there are strong multicollinearity or other numerical problems.



	Country	Both	Male	Female	GDP_CD	DFE
0	Afghanistan	5.7	6.2	5.3	1932.89	33.0
1	Albania	5.9	6.6	5.2	10304.7	41.0
2	Algeria	1.9	2.3	1.5	14193.4	28.0
3	Angola	13.8	20.7	7.3	NaN	12.5
4	Argentina	10.3	17.2	4.1	NaN	34.0





Optimization terminated successfully. Current function value: 0.251107 Iterations 8

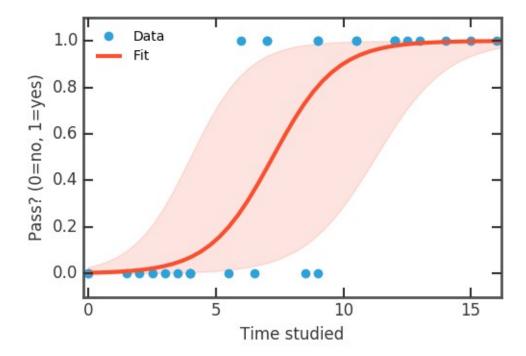
Logit Regression Results

			====						==
Dep. Variable:					Pass	No.	Observations	:	30
Model:				l	ogit	Df R	esiduals:		28
Method:					MLE	Df M	lodel:		1
Date:		Mon,	21	Dec	2015	Pseu	ido R-squ.:	0.63	66
Time:				03:4	10:43	Log-	Likelihood:	-7.53	32
converged:					True	LL-N	ull:	-20.7	28
_						LLR	p-value:	2.791e-	07
		====	====						==
	coef		std	err		Z	P> z	[95.0% Conf. Int	.]
	-5.7980			. 240		2.588	0.010	-10.188 -1.4	
Time	0.8020		0.	. 297		2.703	0.007	0.220 1.3	884

	const	Time
const	5.017663	-0.635081
Time	-0.635081	0.088035

-5.79798670884 0.801979232718

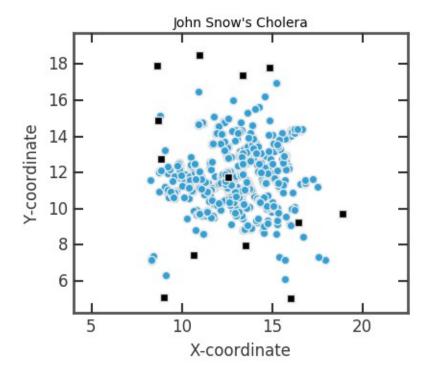
	0	1
const	-10.188333	-1.407640
Time	0.220444	1.383514

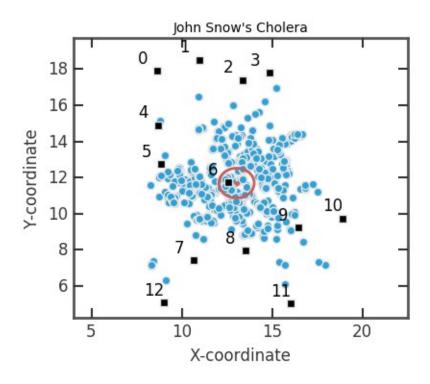


Chapter 5: Clustering

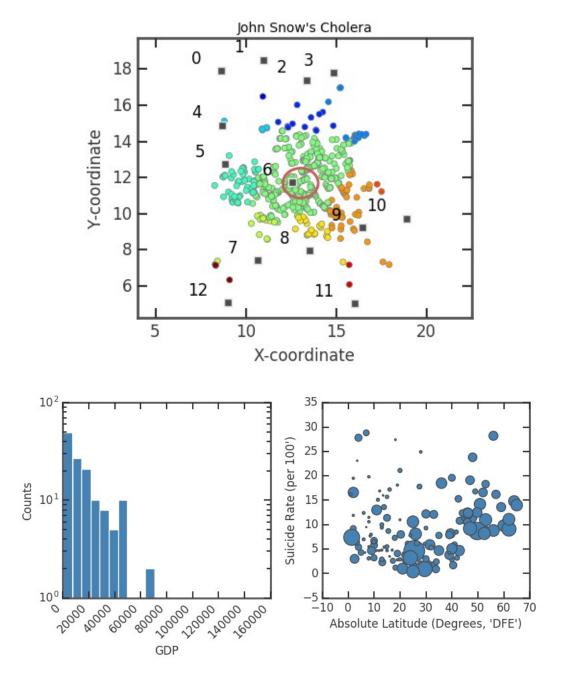
	x	Y
0	13.588010	11.095600
1	9.878124	12.559180
2	14.653980	10.180440
3	15.220570	9.993003
4	13.162650	12.963190

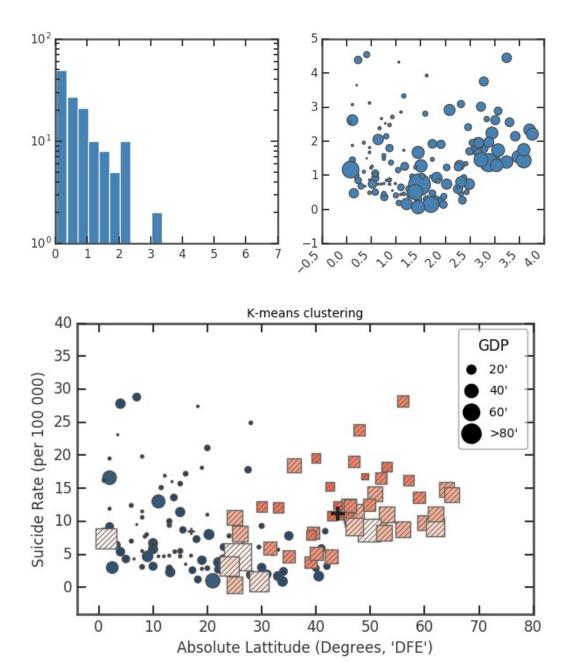
	X	Υ
0	8.651201	17.891600
1	10.984780	18.517851
2	13.378190	17.394541
3	14.879830	17.809919
4	8.694768	14.905470





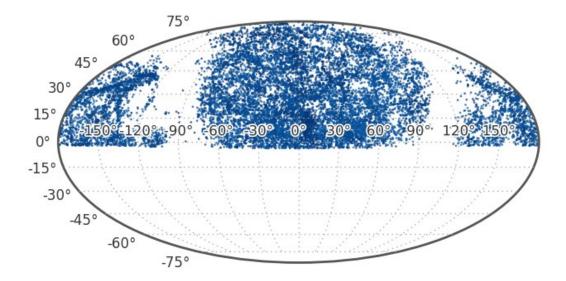
	X	Υ	С
0	13.588010	11.095600	6
1	9.878124	12.559180	5
2	14.653980	10.180440	9
3	15.220570	9.993003	9
4	13.162650	12.963190	6

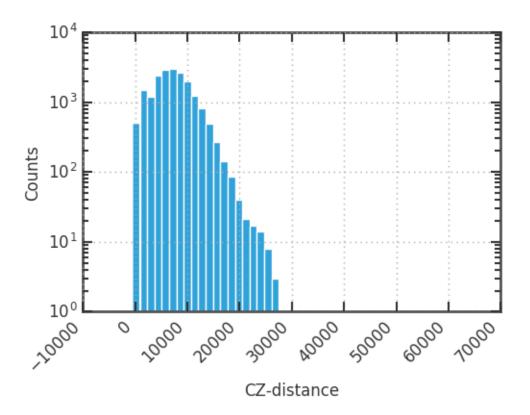


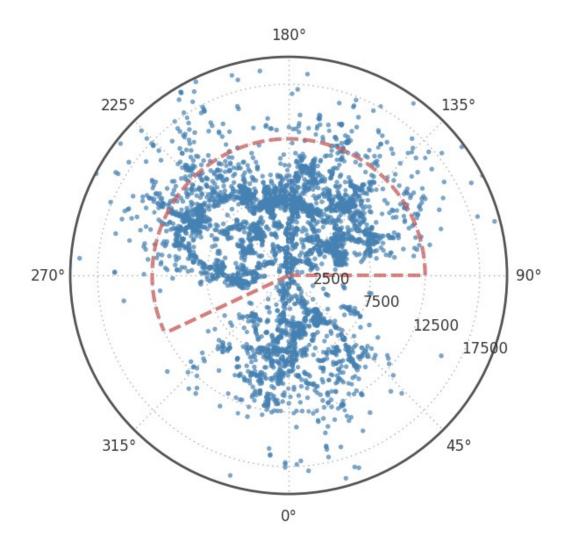


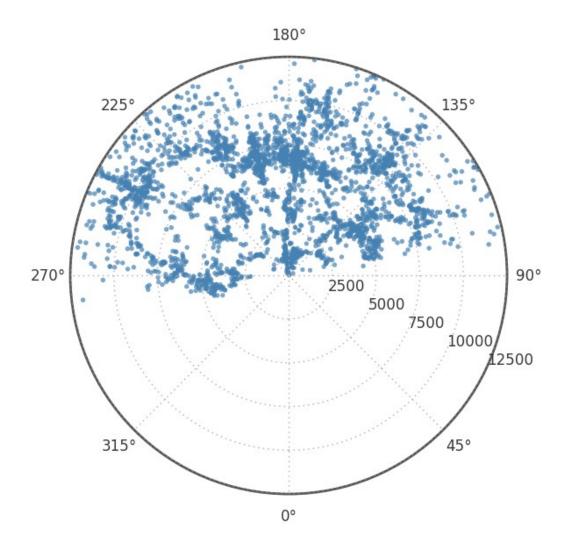
	ra	dec	Zmag	cz	cze	Т	U	Ne	Zname	С	Ref	Oname	M	N
0	000237.9	+163838	14.9	6350	19	Α	1	0	000000+16220	F		15378S		
1	000246.3	+185310	14.8	7864	47	Α	0	0	000012+18370	Z	0650	00002+1837		
2	000257.0	+041231	15.5	8695	40	Ε	0	0	000030+03560	Z	2700	00005+0356		
3	000302.9	+185221	15.5	8007	39	Ε	0	0	000030+18360	Z	0650	00005+1836		
4	000305.6	-015450	14.3	7298	42	В	0	0	000036-02110	Z	2218	00006-0211		

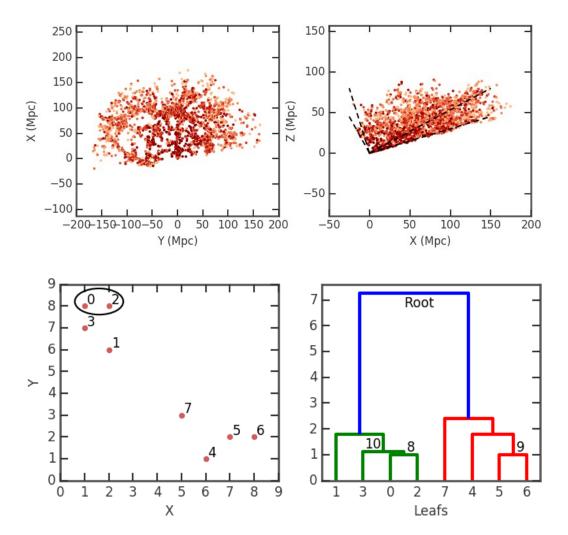
	ra	dec	Zmag	cz	cze	Т	U	Ne	Zname	С	Ref	Oname	M	N
0	00h02m37.9s	+16d38m38s	14.9	6350	19	Α	1	0	000000+16220	F		I5378S		
1	00h02m46.3s	+18d53m10s	14.8	7864	47	Α	0	0	000012+18370	Z	0650	00002+1837		
2	00h02m57.0s	+04d12m31s	15.5	8695	40	Ε	0	0	000030+03560	Z	2700	00005+0356		
3	00h03m02.9s	+18d52m21s	15.5	8007	39	Ε	0	0	000030+18360	Z	0650	00005+1836		
4	00h03m05.6s	-01d54m50s	14.3	7298	42	В	0	0	000036-02110	Z	2218	00006-0211		

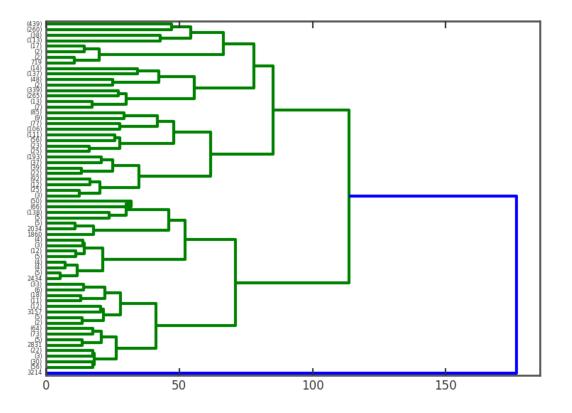


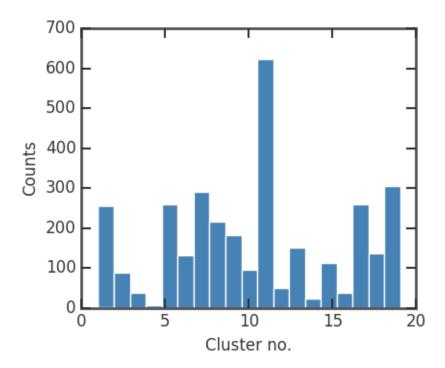


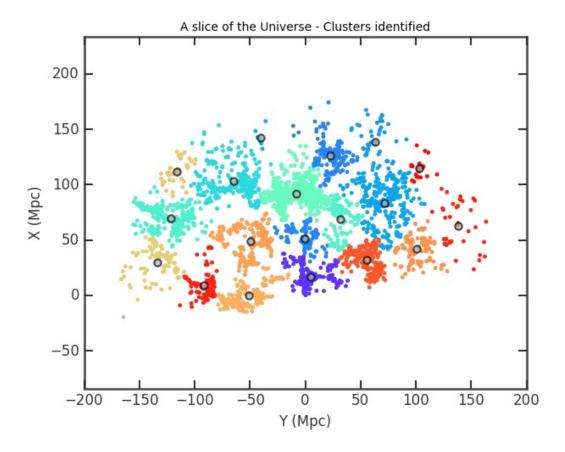




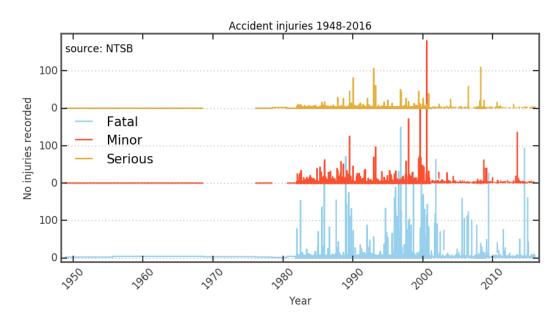


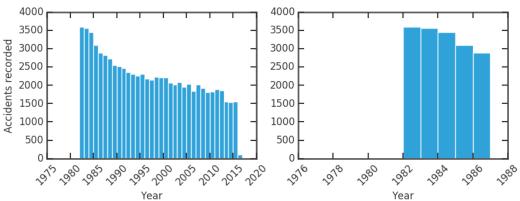


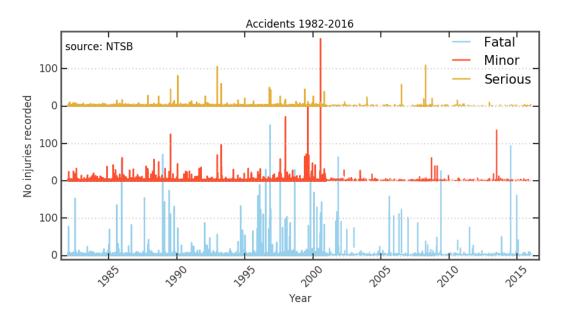




Chapter 6: Bayesian Methods



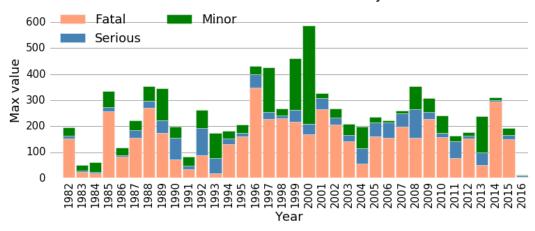




	lat	long	fatal	serious	minor	uninjured	month	year	decyear
1	30.757778	-88.355555	0.443978	0.203699	0.279474	2.317168	6.488450	1982	1982.495213
2	47.080556	-117.368611	0.358996	0.190059	0.296213	4.258810	6.652137	1983	1983.508864
3	NaN	NaN	0.356749	0.202322	0.303919	3.621739	6.553659	1984	1984.502076
4	NaN	NaN	0.534198	0.198379	0.359507	3.663855	6.477390	1985	1985.494267
5	NaN	NaN	0.410435	0.215454	0.338097	4.138531	6.480556	1986	1986.495181

```
1984.,
                                 1985.,
                                         1986.,
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                1983.,
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                                                                   1989.,
array([ 1982.,
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                                                                   2013.,
        2014.,
                2015.,
                         2016.])
```

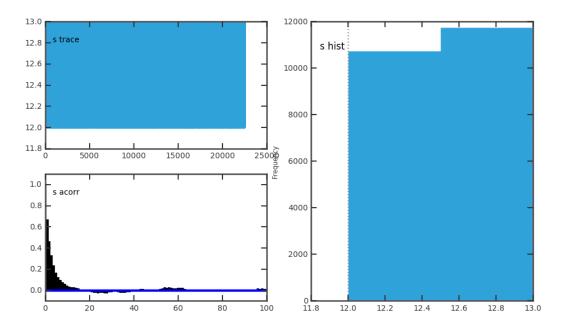
Plane accidents maximum injuries

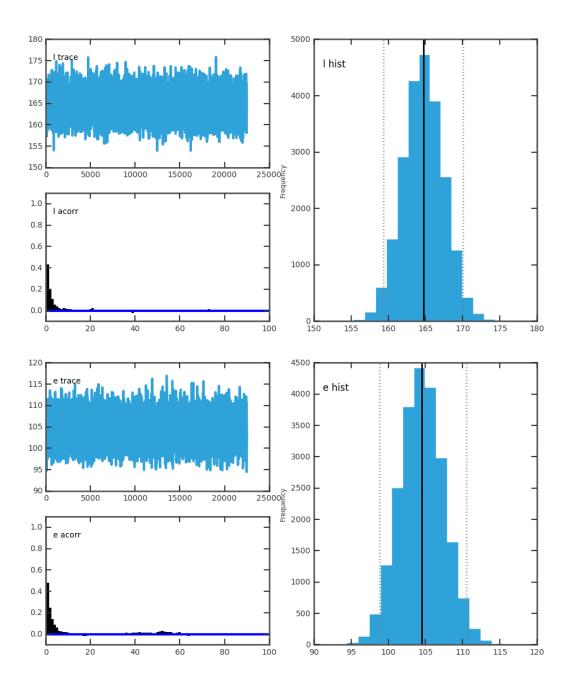


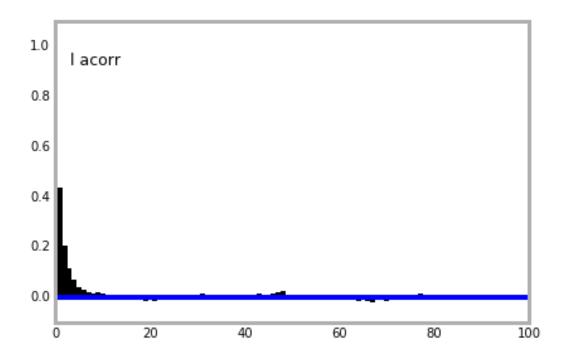
--] 50000 of 50000 complete in 8.6 sec

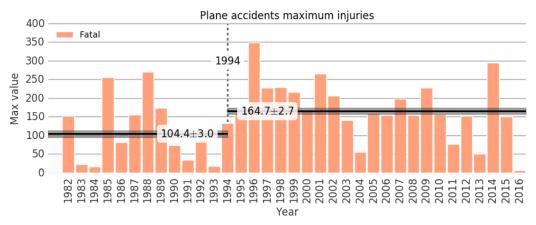
 ${<pymc.distributions.new_dist_class.<locals>.new_class 'l' at 0x7f841e56b470>: [<pymc.StepMethods.Metropolis at 0x7f841e56b390>],}$ 841e56b399>],

<pre



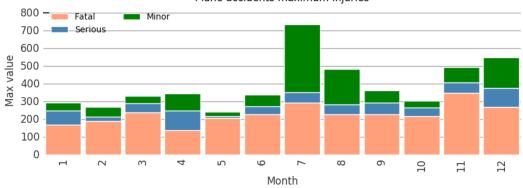


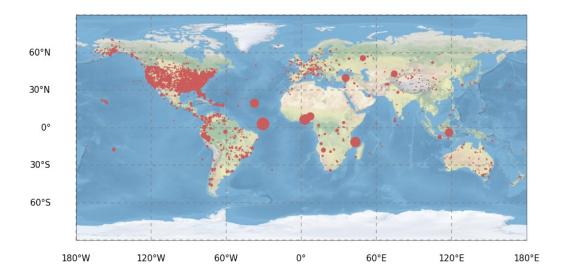


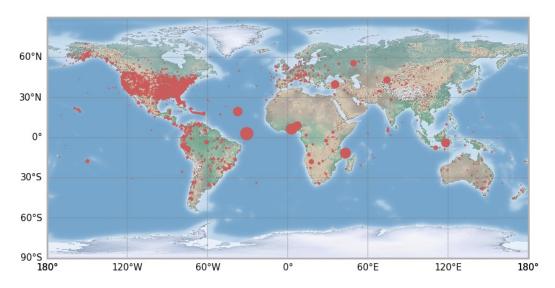


| | lat | long | fatal | serious | minor | uninjured | month | year | decyear |
|---|-----------|------------|----------|----------|----------|-----------|-------|-------------|-------------|
| 1 | 35.355070 | -91.799283 | 1.024984 | 0.351452 | 0.482369 | 7.505215 | 1 | 1996.595841 | 1996.637186 |
| 2 | 35.211711 | -92.921433 | 0.890277 | 0.295262 | 0.486586 | 6.861646 | 2 | 1996.190641 | 1996.313377 |
| 3 | 36.439443 | -93.719725 | 0.673851 | 0.281289 | 0.428362 | 6.491056 | 3 | 1996.299847 | 1996.503531 |
| 4 | 37.187038 | -93.330245 | 0.651183 | 0.304870 | 0.489868 | 5.455664 | 4 | 1996.292600 | 1996.579883 |
| 5 | 37.816921 | -94.954229 | 0.671199 | 0.272957 | 0.445699 | 5.194848 | 5 | 1996.382683 | 1996.754983 |

Plane accidents maximum injuries



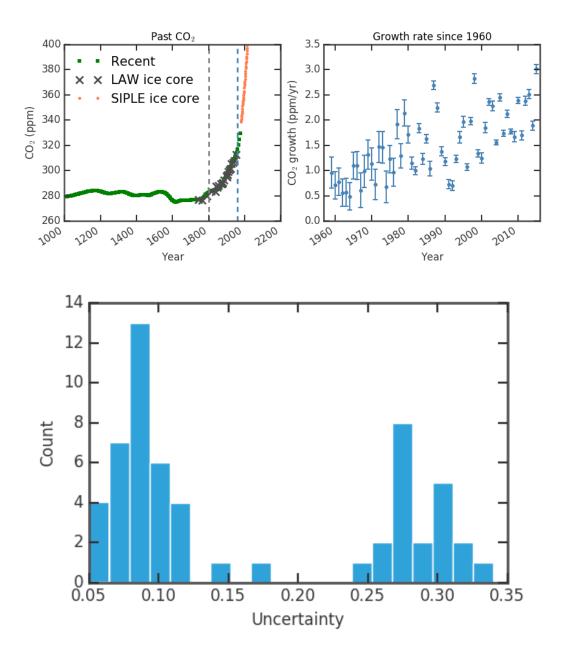


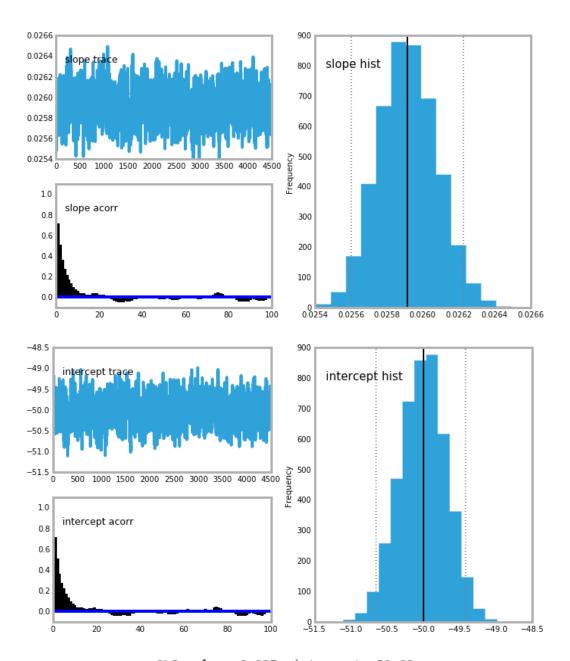


| | depth | year | co2 |
|----|-------|----------------|-----------|
| 20 | 86.80 | 1943 | 307.9 |
| 21 | 81.22 | 1953 | 312.7 |
| 22 | Data | in | the |
| 23 | table | were | published |
| 24 | CO2 | concentrations | are |

object float64 float64 float64

(dtype('float64'), dtype('int64'))

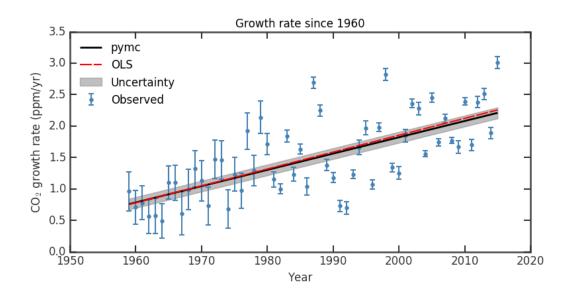




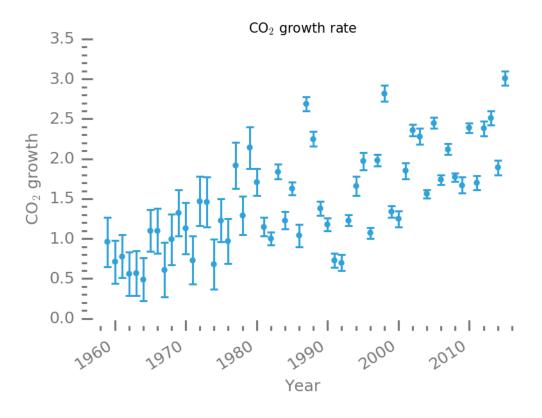
OLS: slope:0.027, intercept:-51.81 Bay: slope:0.026, intercept:-50.01

| | 0 | 1 |
|-----------|------------|------------|
| Intercept | -66.531103 | -37.092365 |
| year | 0.019425 | 0.034240 |

```
{'intercept': {'95% HPD interval': array([-50.64990877, -49.42168758]),
  'mc error': 0.011931753405578262,
  'mean': -50.008066085637815,
  'n': 4500,
  'quantiles': {2.5: -50.646760270081515,
  25: -50.228993605017173,
  50: -50.004119841092979,
  75: -49.789140434449351,
  97.5: -49.412790669964195},
  'standard deviation': 0.31970980437231761},
 'slope': {'95% HPD interval': array([ 0.02560229,  0.02622579]),
  'mc error': 6.0141748688389682e-06,
  'mean': 0.0259157602742307,
  'n': 4500,
  'quantiles': {2.5: 0.025609913276748567,
  25: 0.025803373826353462,
  50: 0.02591258446734945,
  75: 0.026027359901668552,
  97.5: 0.026237852465415799},
  'standard deviation': 0.00016186303140381465}}
```



Chapter 7: Supervised and Unsupervised Learning

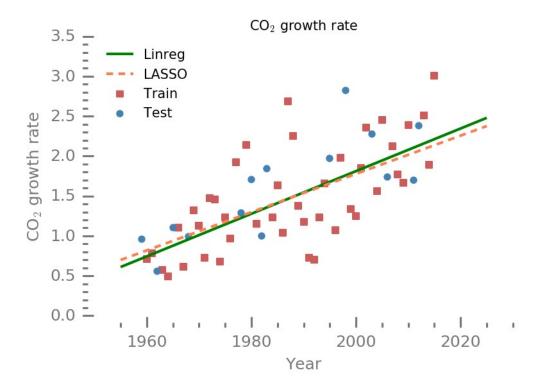


+++++ LinearRegression ++++++
Slope: 0.027 Intercept:-51.60
Mean squared residuals: 0.17

Variance score: 0.56

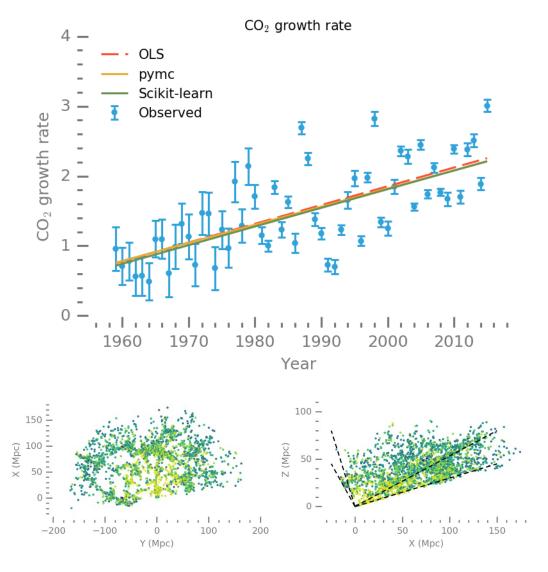
Slope: 0.024 Intercept:-46.16 Mean squared residuals: 0.17

Variance score: 0.56

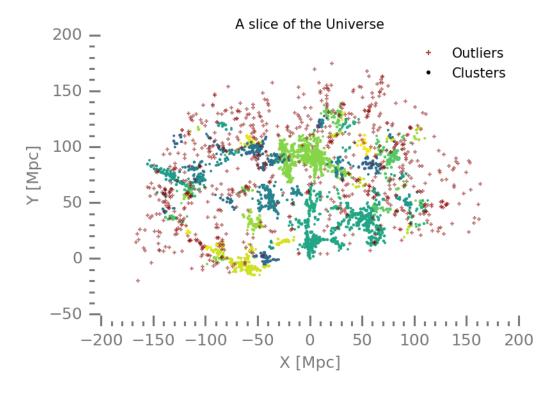


LinearSVC: 0.49 LASSO: 0.48

Slope Intercept ML: 0.027 -51.597 OLS: 0.027 -51.812 Bay: 0.026 -49.998

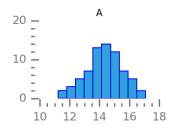


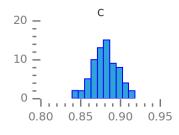
Estimated number of clusters: 8 Silhouette Coefficient: -0.143

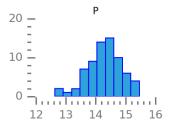


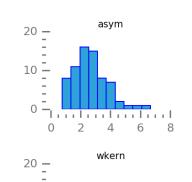
- 1. area A,
- 2. perimeter P,
- 3. compactness $C = 4*pi*A/P^2$,
- 4. length of kernel,
- 5. width of kernel,
- asymmetry coefficient
- 7. length of kernel groove.
- 8. group

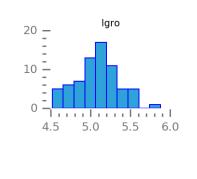
| | Α | Р | С | Ikern | wkern | asym | Igro | gr |
|---|-------|-------|--------|-------|-------|-------|-------|----|
| 0 | 15.26 | 14.84 | 0.8710 | 5.763 | 3.312 | 2.221 | 5.220 | 1 |
| 1 | 14.88 | 14.57 | 0.8811 | 5.554 | 3.333 | 1.018 | 4.956 | 1 |
| 2 | 14.29 | 14.09 | 0.9050 | 5.291 | 3.337 | 2.699 | 4.825 | 1 |
| 3 | 13.84 | 13.94 | 0.8955 | 5.324 | 3.379 | 2.259 | 4.805 | 1 |
| 4 | 16.14 | 14.99 | 0.9034 | 5.658 | 3.562 | 1.355 | 5.175 | 1 |

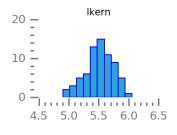


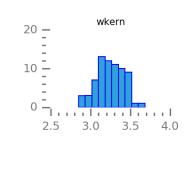


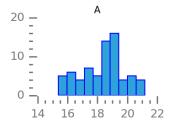


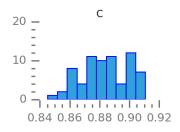


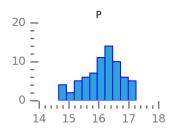


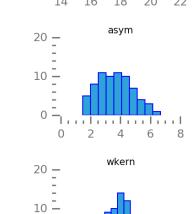


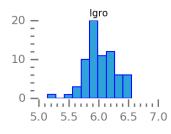


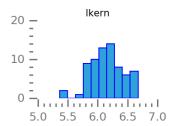


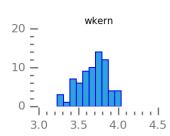


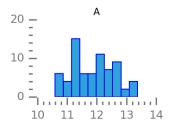


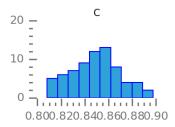


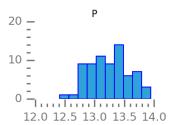


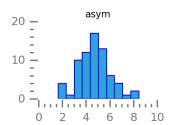


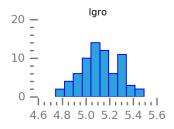


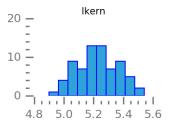


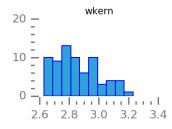


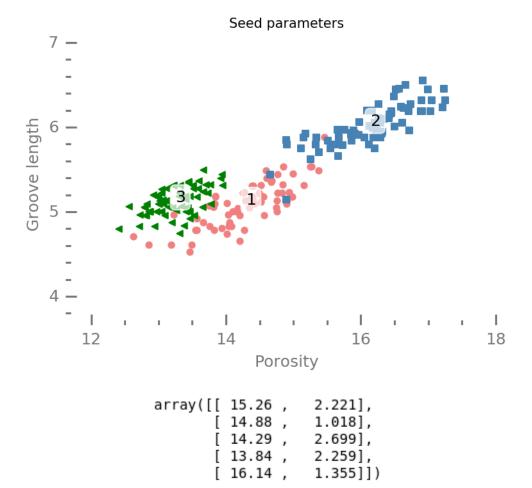






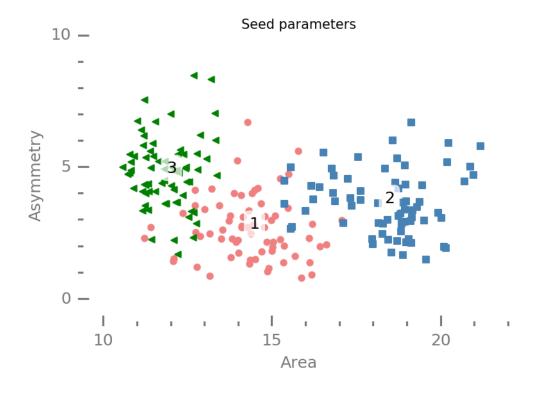


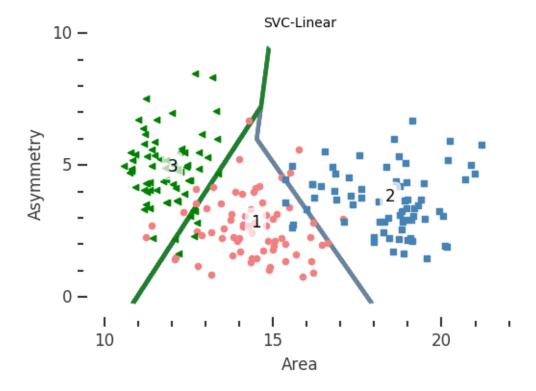


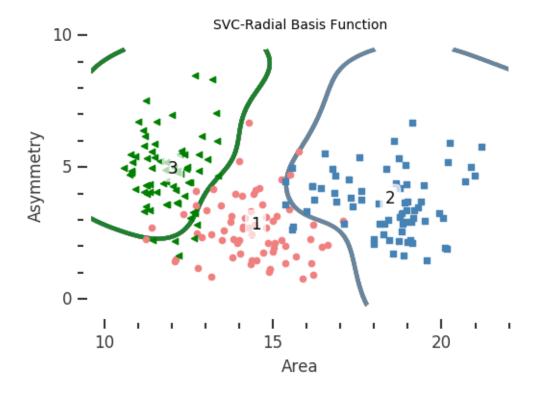


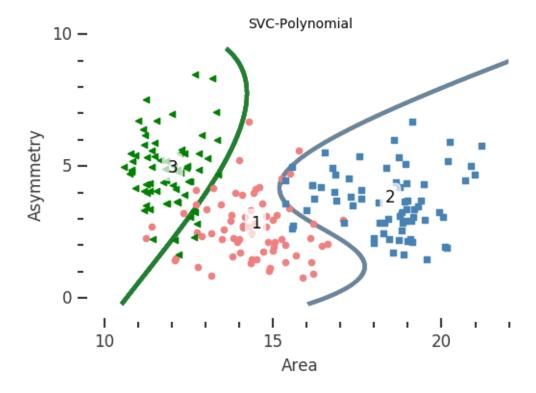
| | Α | Р | С | lkern | wkern | asym | Igro | gr |
|---|-------|-------|--------|-------|-------|-------|-------|----|
| 0 | 15.26 | 14.84 | 0.8710 | 5.763 | 3.312 | 2.221 | 5.220 | 1 |
| 1 | 14.88 | 14.57 | 0.8811 | 5.554 | 3.333 | 1.018 | 4.956 | 1 |
| 2 | 14.29 | 14.09 | 0.9050 | 5.291 | 3.337 | 2.699 | 4.825 | 1 |
| 3 | 13.84 | 13.94 | 0.8955 | 5.324 | 3.379 | 2.259 | 4.805 | 1 |
| 4 | 16.14 | 14.99 | 0.9034 | 5.658 | 3.562 | 1.355 | 5.175 | 1 |

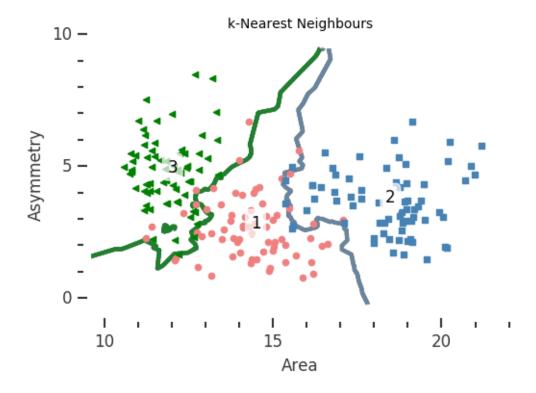
1.355]])

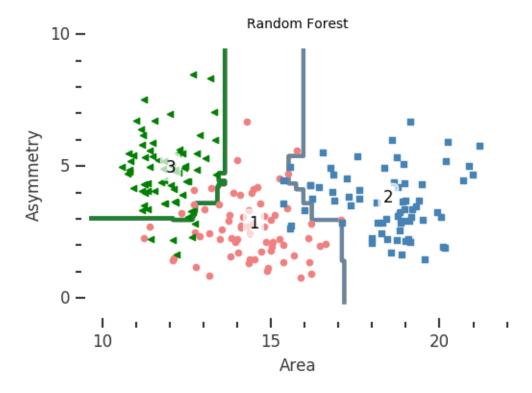






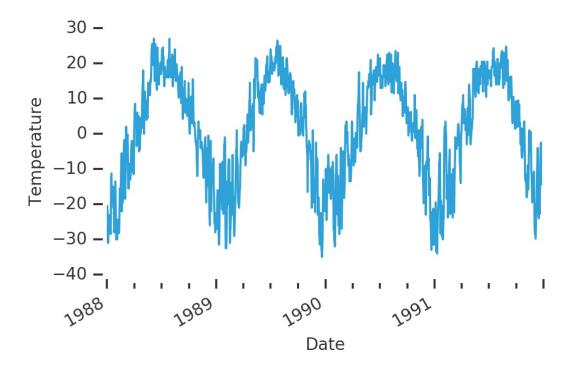






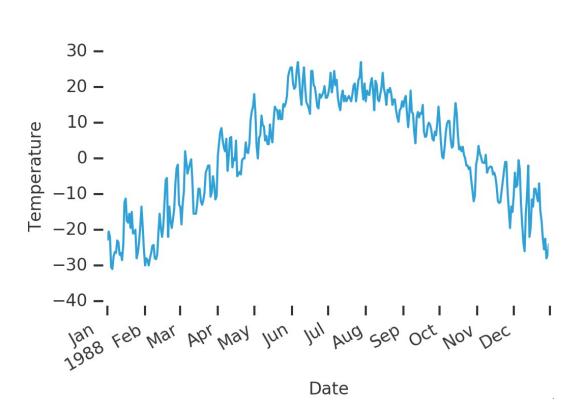
Chapter 8: Time Series Analysis

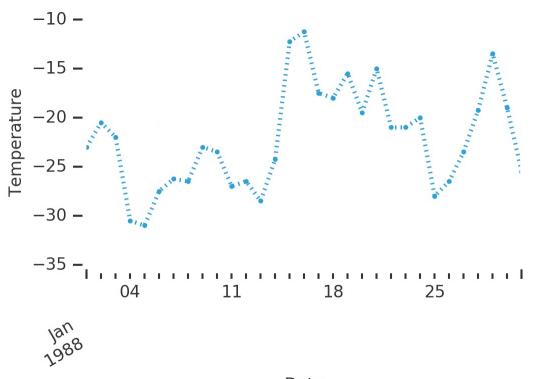
| | Temp |
|------------|-------|
| Date | |
| 1988-01-01 | -23.0 |
| 1988-01-02 | -20.5 |
| 1988-01-03 | -22.0 |
| 1988-01-04 | -30.5 |
| 1988-01-05 | -31.0 |



1461.000000 count 0.803320 mean std 15.154634 min -35.000000 25% -11.250000 50% 2.000000 75% 14.500000 27.000000 max

Name: Temp, dtype: float64

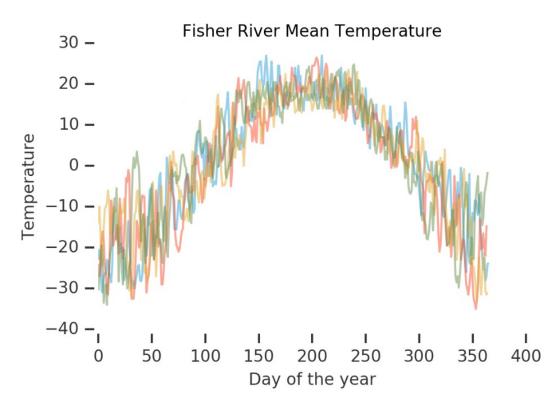




Date

Date 1988-01-04 -30.50 1988-01-05 -31.00 1988-01-06 -27.50 1988-01-07 -26.25 1988-01-08 -26.50

Name: Temp, dtype: float64



Date

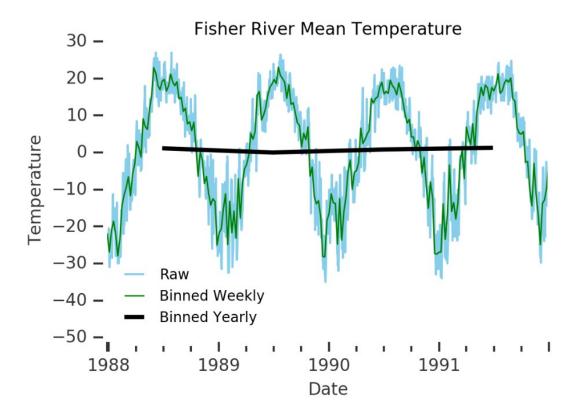
1988-12-31 1.138661

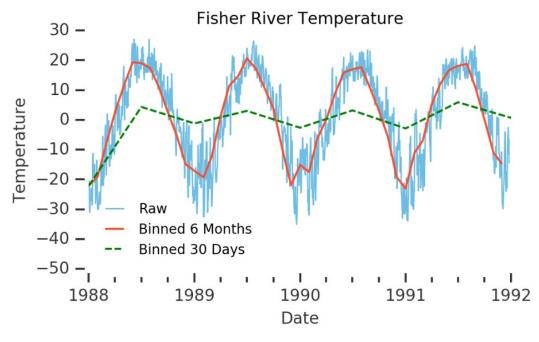
1989-12-31 -0.006164

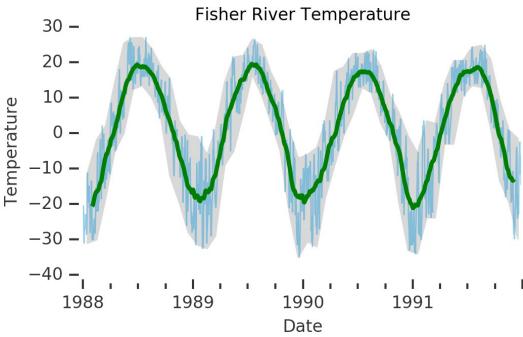
1990-12-31 0.815753

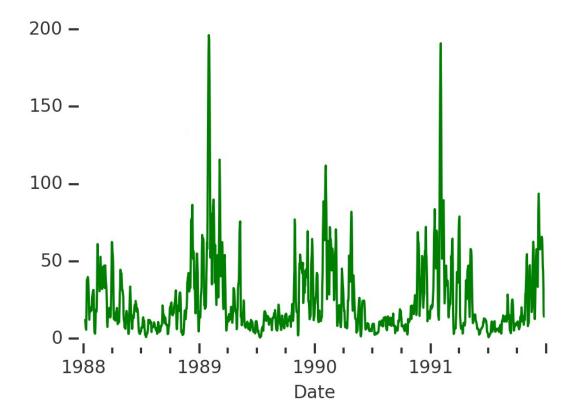
1991-12-31 1.264110

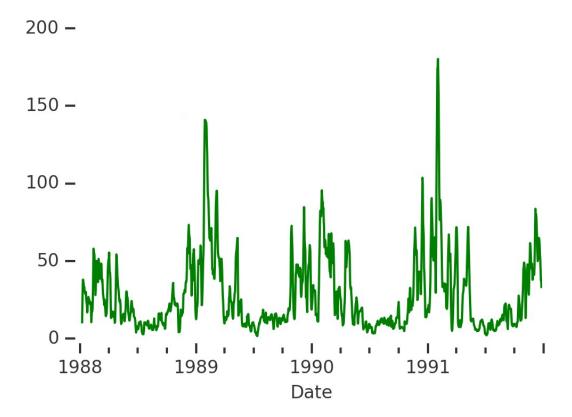
Freq: A-DEC, Name: Temp, dtype: float64

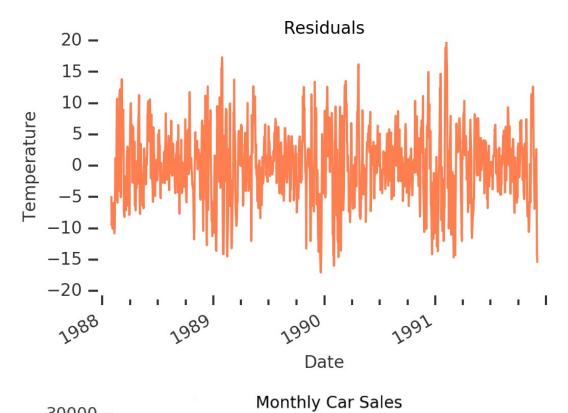


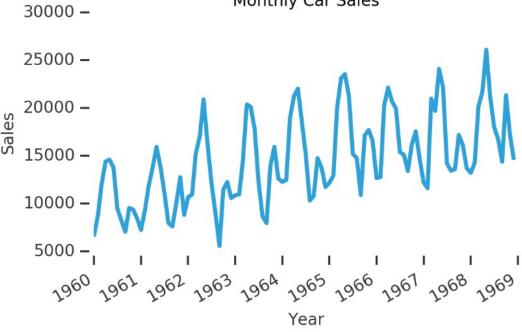










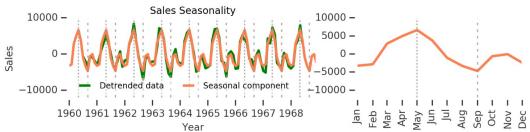


adf -1.673 p 0.763

crit. val. 1%: -4.060, 5%: -3.459, 10%: -3.155

stationary? false

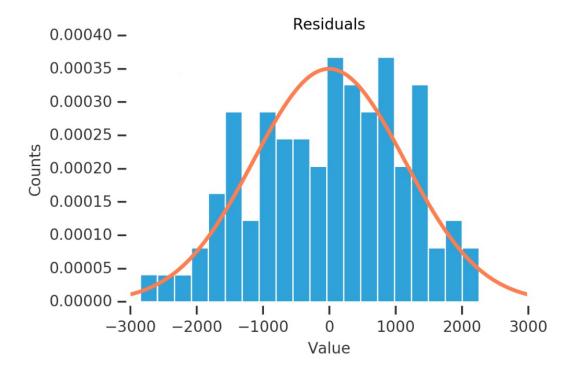


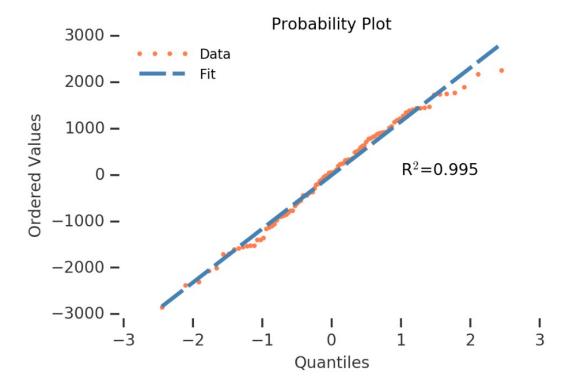


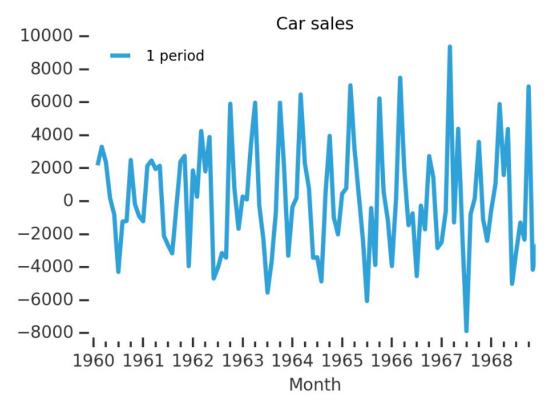
adf -4.501 p 0.0015

crit. val. 1%: -4.072, 5%: -3.465, 10%: -3.159

stationary? true



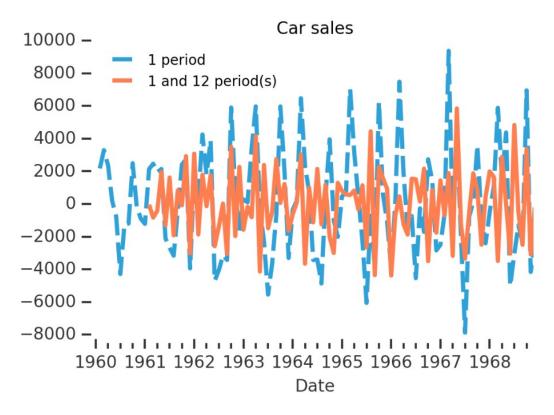




adf -3.124 p 0.101

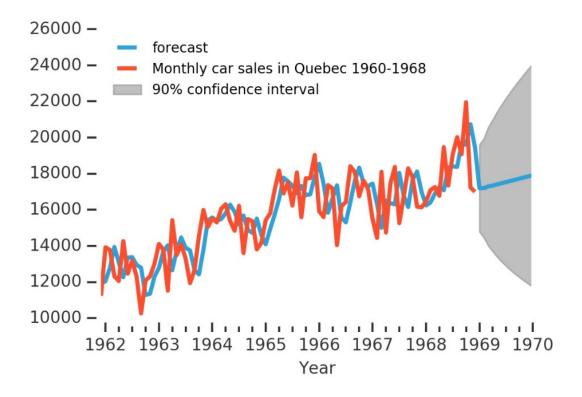
crit. val. 1%: -4.061, 5%: -3.459, 10%: -3.156

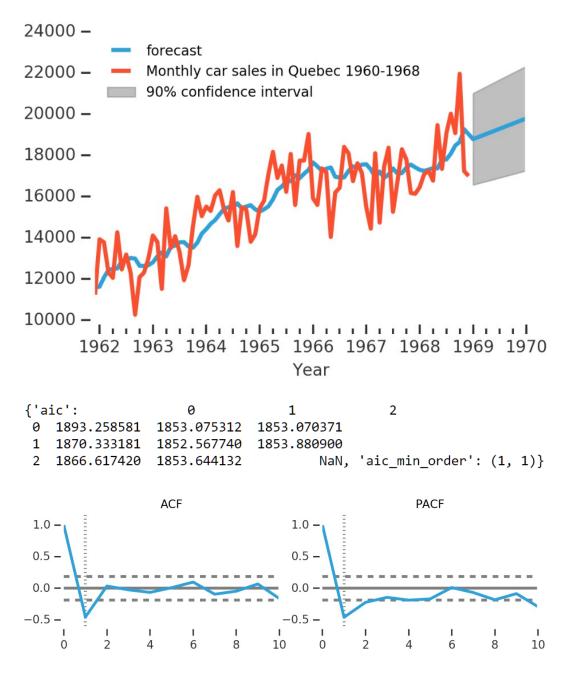
stationary? false

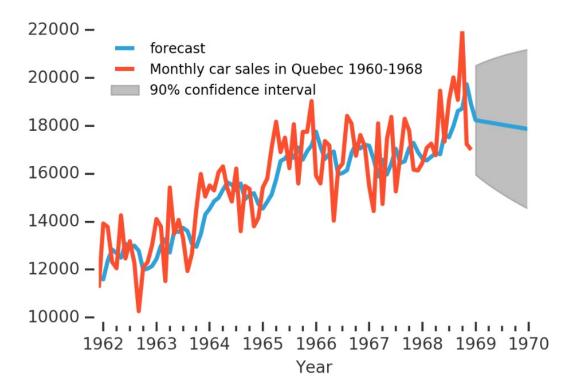


adf -3.875
p 0.0131
crit. val. 1%: -4.077, 5%: -3.467, 10%: -3.160
stationary? true

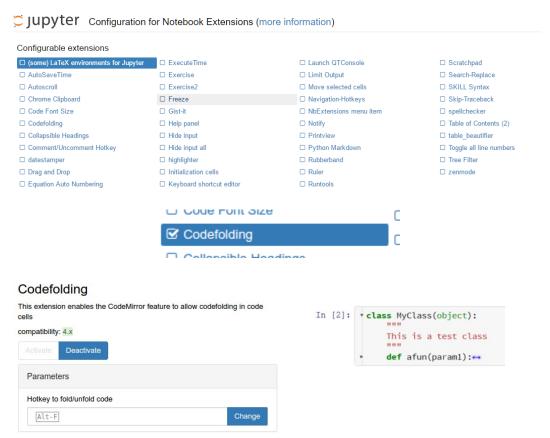
adf -3.611
p 0.0289
crit. val. 1%: -4.061, 5%: -3.459, 10%: -3.156
stationary? true







Appendix: More on Jupyter Notebook and matplotlib Styles



/nbextensions/usability/codefolding/readme.md?v=20160501205159

This extension adds codefolding functionality from CodeMirror to a codecell.

After clicking on the gutter (left margin of codecell) or typing Alt+F, the code gets folded. See the examples below. The folding status is saved in the cell metadata of the notebook, so reloading of a notebook will restore the folding view.

Supported modes

Three different folding modes are supported:

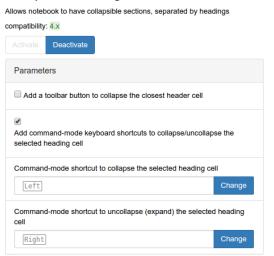
```
def function_one(p1,p2):
    sel = [p1, p2][np.random.randint(2)]
    return sel

def function_one(p1,p2):

# it is possible to use comments
# to hide code as well
function_one(10,30)
```

▶ # it is possible to use comments ↔

Collapsible Headings





1 A report

This document gives short examples on how to work with the Jupyter Notebook and some of the extensions that I hope that you have now activated. It also shows that you can actually write nice summaries of the analysis, perhaps not for the final version of a report, but for early drafts it should be ideal.

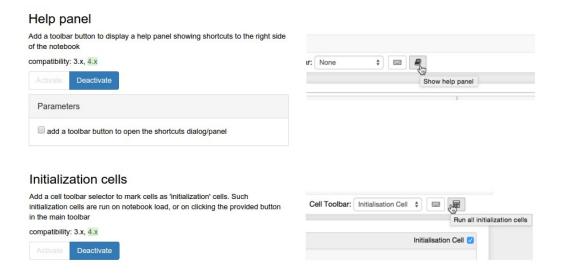
▼ 1.1 Some example text and code

In a Markdown cell you can for example

- · Create headings
- · Type normal text, just like any text editor and

▶ 1 A report

[...]





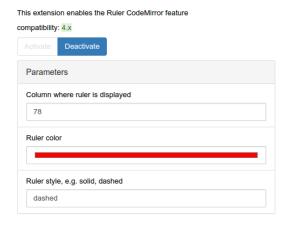
NbExtensions menu item

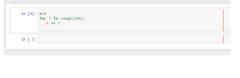
Add an edit-menu item to open the NbExtensions config page compatibility: 4.x

Activate Deactivate



Ruler





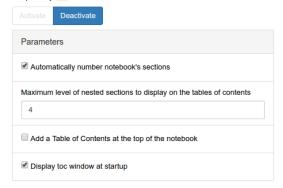
%matplotlib inline
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

```
A
                                                                      values = (1+np.array([0, 1.e-15]))*1.e27
    In [33]:
              plt.plot(values)
            C:\Users\\___\Anaconda3\lib\site-packages\matplotlib\ticker.py in set form
            at(self, vmin, vmax)
                608
                           thresh = 1e-3 * 10 ** loc_range_oom
                609
                           while sigfigs >= 0:
                               if np.abs(locs - np.round(locs, decimals=sigfigs)).max()
            --> 610
            < thresh:
                611
                                  sigfigs -= 1
                612
                               else:
            C:\Users\\Anaconda3\lib\site-packages\numpy\core\fromnumeric.py in roun
            d (a, decimals, out)
               2791
                       except AttributeError:
                          return _wrapit(a, 'round', decimals, out)
               2792
            -> 2793
                       return round(decimals, out)
               2794
               2795
            AttributeError: 'float' object has no attribute 'rint'
CellToolbar
                           N ■ C Code
                                                   (See See
                                                                      =
               values = (1+np.array([0, 1.e-15]))*1.e27
    In [34]:
               plt.plot(values)
    Out[34]: [<matplotlib.lines.Line2D at 0x2910e536128>]
             Error in callback <function install_repl_displayhook.<locals>.post_execute at 0
             x00000291040B30D0> (for post_execute):
             AttributeError: 'float' object has no attribute 'rint'
             AttributeError: 'float' object has no attribute 'rint'
             <matplotlib.figure.Figure at 0x2910e405828>
```

Table of Contents (2)

The ToC2 extension displays a floating (draggable) table of contents of the notebooks headers. Optionally, it also allows to automatically number all notebook's sections, and to add a table of Contents cell at the top of the notebook.

compatibility: 4.x







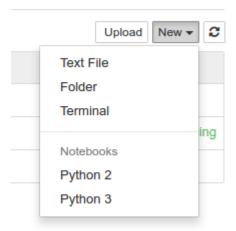
Contents [-] On t

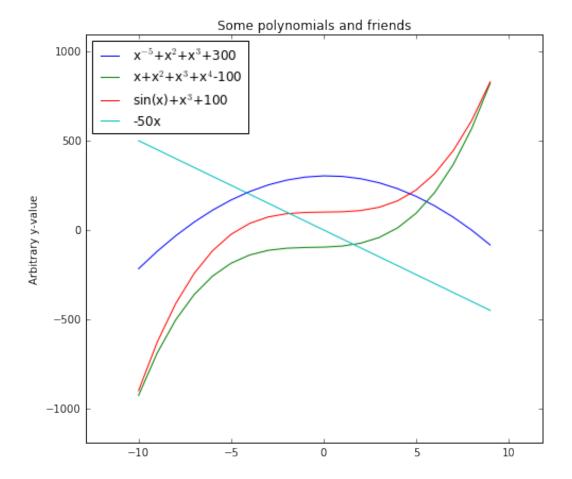
- 1 A report
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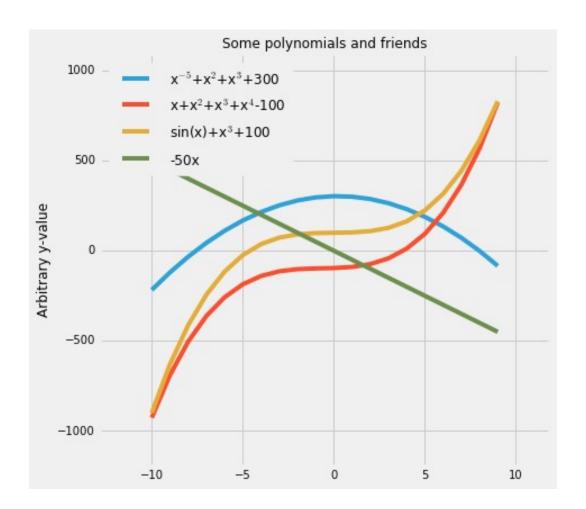
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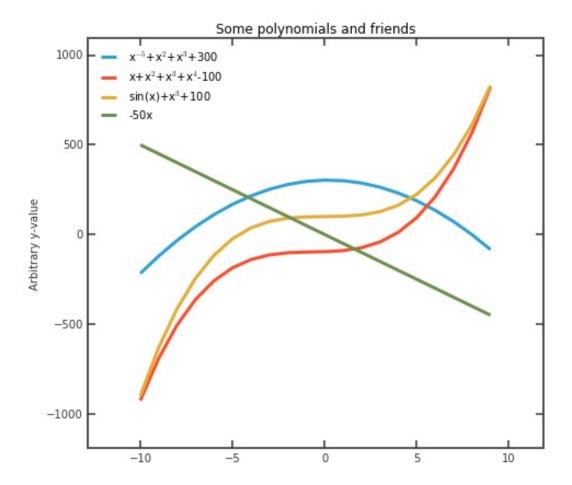
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n [1]: import warnings









Some polynomials and friends

