

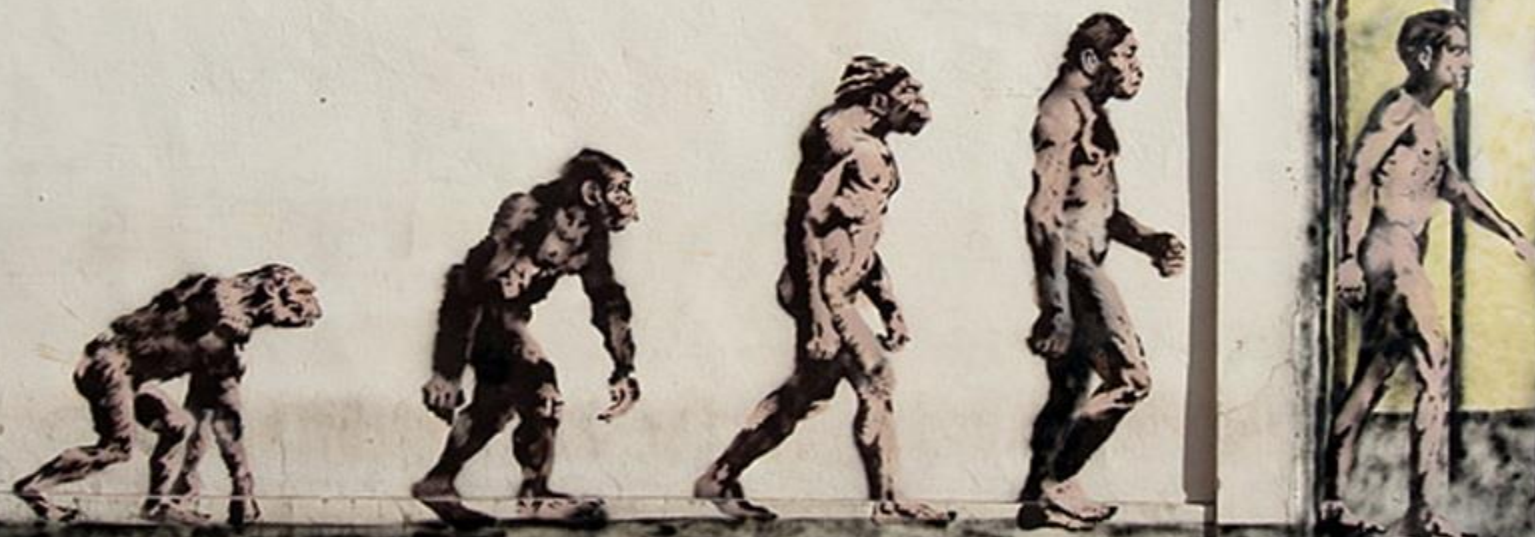
An introduction to Jupyter Notebooks for Social Science Research

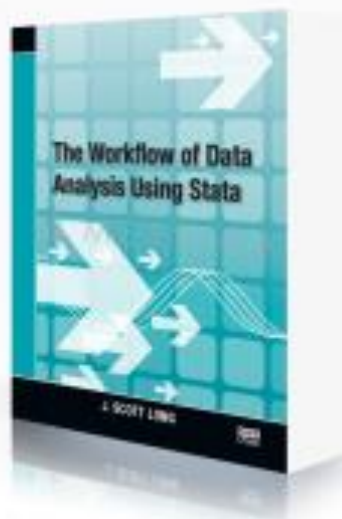
Vernon Gale

University of Edinburgh

CARPENTERS PLACE

NCRM







As with many scientists, Linus Pauling utilized bound notebooks to keep track of the details of his research as it unfolded. A testament to the remarkable length and diversity of Dr. Pauling's career, the Pauling Papers holdings include forty-six research notebooks spanning the years of 1922 to 1994 and covering any number of the scientific fields in which Dr. Pauling involved himself. In this regard, the notebooks contain many of Pauling's laboratory calculations and experimental data, as well as scientific conclusions, ideas for further research and numerous autobiographical musings.

[Research Notebook 01](#)

1922

[Research Notebook 02](#)

1922-1923, 1932, 1934, 1936, 1973,
1985

[Research Notebook 03](#)

1923-1925

[Research Notebook 04](#)

1923-1924, 1928-1930

[Research Notebook 05](#)

[Research Notebook 13](#)

1935-1936, 1938-1939

[Research Notebook 14](#)

1936-1939, 1949, 1952

[Research Notebook 15](#)

1935, 1937, 1968

[Research Notebook 16](#)

1935-1956

[Research Notebook 17](#)

1939-1941, 1971, 1988

[Research Notebook 24](#)

1953, 1956, 1962, 1963, 1967, 1968,
1969, 1970, 1973

[Research Notebook 25](#)

1958, 1964-1966

[Research Notebook 26](#)

1955, 1964-1969, 1974-1976, 1980-
1982, 1987, 1990-1991

[Research Notebook 27](#)

1952-1954, 1960-1961, 1964, 1971-

[Research Notebook 35b](#)

1938-1939, 1946, 1955, 1968, 1986-
1988

[Research Notebook 36](#)

1980-1981, 1986-1987

[Research Notebook 37](#)

1971, 1983

[Research Notebook 38](#)

1980-1981, 1983, 1985, 1989

[Research Notebook 39](#)

But on January 10th the stars appeared in the following position with regard to Jupiter; there were two only, and both on the east side

Ori.



Occ.

of Jupiter, the third, as I thought, being hidden by the planet.

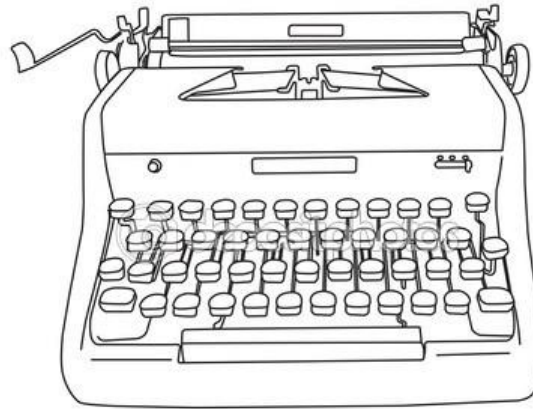




<https://www.youtube.com/watch?v=BmHPoBpZoJ4>



Documentation; Portability; Language agnostic; Rich visual outputs; Big data tools; Teaching tools; Collaboration





In []:

```
In [4]: summarize
```

```
In [4]: summarize
```

Variable	Obs	Mean	Std. Dev.	Min	Max
-----+-----					
case	1,580	517.7411	284.8605	1	1003
femp	1,580	.6455696	.4784918	0	1
mune	1,580	.0740506	.2619362	0	1
time	1,580	7.2	3.981019	0	13
und1	1,580	.0746835	.2629633	0	1
-----+-----					
und5	1,580	.2974684	.4572891	0	1
age	1,580	36.01013	9.114841	18	60

```
In [4]: summarize
```

```
-----+-----  
case |      1,580      517.7411      284.8605          1      1003  
femp |      1,580      .6455696      .4784918          0          1  
mune |      1,580      .0740506      .2619362          0          1  
time |      1,580          7.2      3.981019          0          13  
und1 |      1,580      .0746835      .2629633          0          1  
-----+-----  
und5 |      1,580      .2974684      .4572891          0          1  
age |      1,580      36.01013      9.114841         18          60
```

The data mirror a real example of data analysed in Davies et al. (1992).

The dataset is a panel of 155 married women.

Davies, Richard B., Peter Elias, and Roger Penn. "The relationship between a husband's unemployment and his wife's participation in the labour force." *Oxford Bulletin of Economics and Statistics* 54.2 (1992): 145-171.

```
In [11]: logit femp mune und5
```

```
Iteration 0:  log likelihood = -1027.2309
Iteration 1:  log likelihood = -879.88806
Iteration 2:  log likelihood = -878.68101
Iteration 3:  log likelihood = -878.67998
Iteration 4:  log likelihood = -878.67998
```

```
Logistic regression              Number of obs   =      1,580
                                LR chi2(2)       =      297.10
                                Prob > chi2      =      0.0000
Log likelihood = -878.67998      Pseudo R2      =      0.1446
```

```
-----+-----
      femp |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      mune | -1.703308   .2358489    -7.22   0.000   -2.165563   -1.241053
      und5 | -1.733521   .1221909   -14.19   0.000   -1.973011   -1.494031
      _cons |  1.306829   .0744154    17.56   0.000    1.160978    1.452681
-----+-----
```

```
In [3]: mylogit <- glm(femp ~ mune + und5, data = mydata, family = "binomial")  
  
summary(mylogit)
```

Call:

```
glm(formula = femp ~ mune + und5, family = "binomial", data = mydata)
```

Deviance Residuals:

	Min	1Q	Median	3Q	Max
	-1.7586	-1.0024	0.6922	0.6922	2.1177

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	1.30683	0.07442	17.561	< 2e-16 ***
mune	-1.70331	0.23585	-7.222	5.12e-13 ***
und5	-1.73352	0.12219	-14.187	< 2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 2054.5 on 1579 degrees of freedom
Residual deviance: 1757.4 on 1577 degrees of freedom
AIC: 1763.4


```
In [6]: independentVar = ['mune', 'und5', 'Int']
logReg = sm.Logit(df['femp'], df[independentVar])
answer = logReg.fit()
```

```
Optimization terminated successfully.
      Current function value: 0.556127
      Iterations 5
```

the results are in the object "answer"

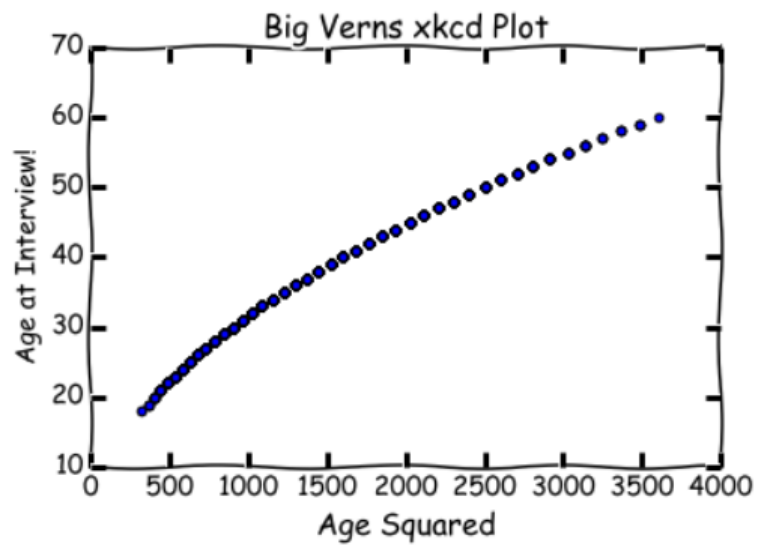
```
In [9]: answer.summary()
```

Out[9]:

Logit Regression Results

Dep. Variable:	femp	No. Observations:	1580
Model:	Logit	Df Residuals:	1577
Method:	MLE	Df Model:	2
Date:	Fri, 14 Oct 2016	Pseudo R-squ.:	0.1446
Time:	10:13:23	Log-Likelihood:	-878.68
converged:	True	LL-Null:	-1027.2
		LLR p-value:	3.056e-65

	coef	std err	z	P> z	[95.0% Conf. Int.]
mune	-1.7033	0.236	-7.222	0.000	-2.166 -1.241
und5	-1.7335	0.122	-14.187	0.000	-1.973 -1.494
Int	1.3068	0.074	17.561	0.000	1.161 1.453

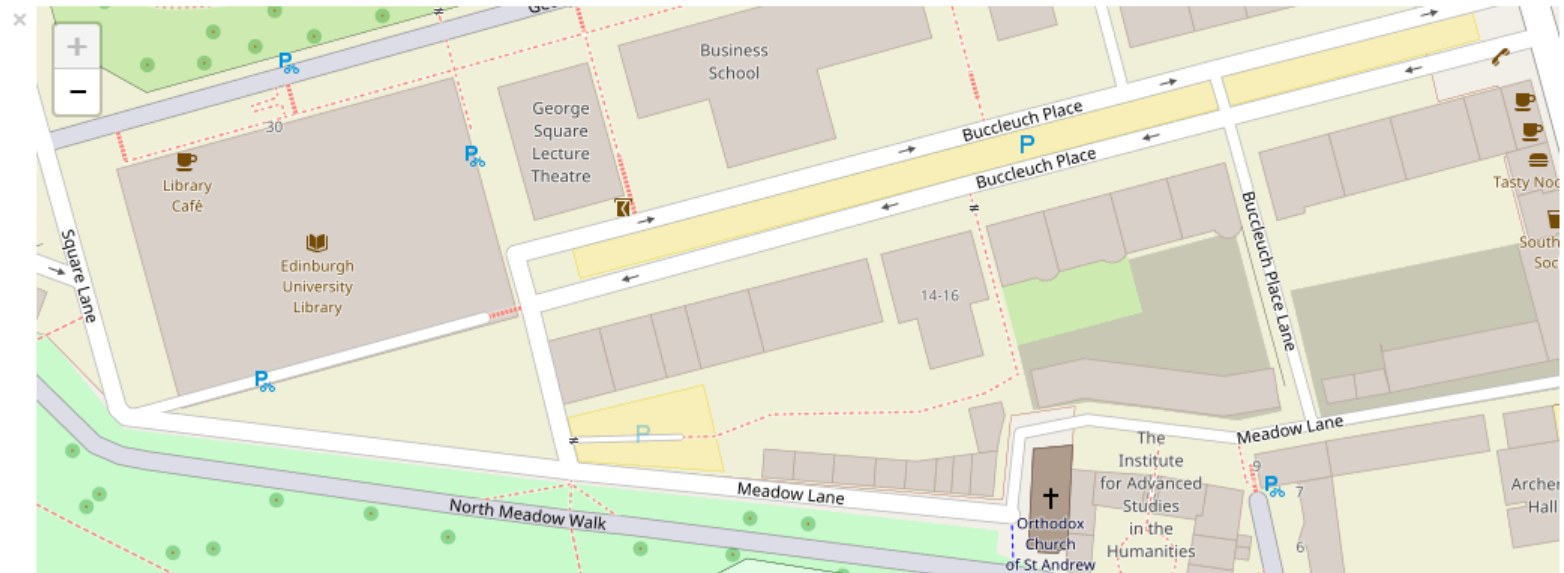


Another inventive use of the wemp dataset

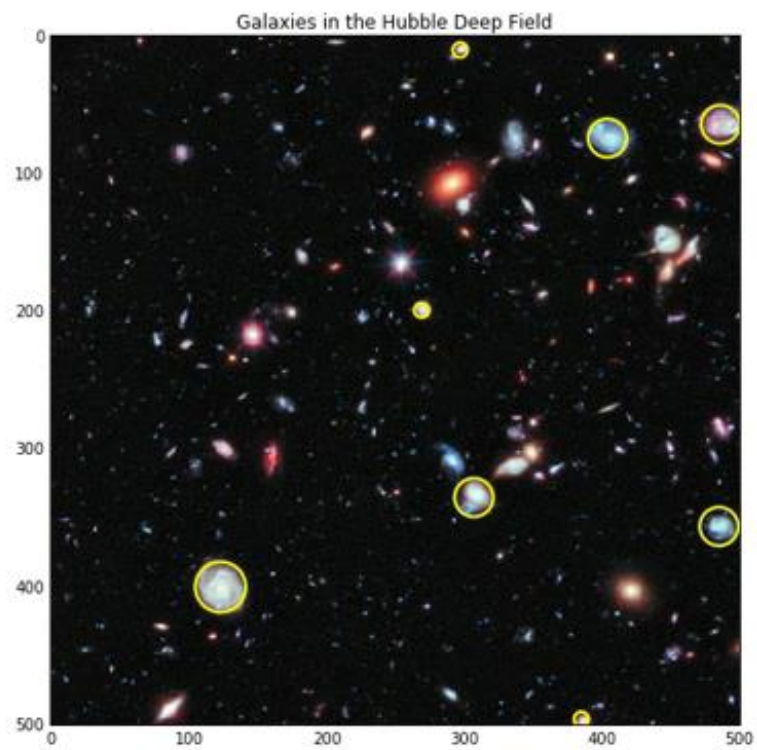
Using an open street map

I've recently moved to a more commodious office in Buccleuch Place. Here is an example of an open source map on my new hood.

```
In [5]: from ipyleaflet import Map
Map(center=[55.942535, -3.187269], zoom=20)
```



× max_sigm 40
a
threshold 0.02
☐ gray



Lorena A. Barba group




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CODE



Prof. Barba awarded a 2016 Leamer-Rosenthal Prize for Open Social Science



The 2016 Leamer-Rosenthal Prizes were announced on 15 December 2016, at the

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