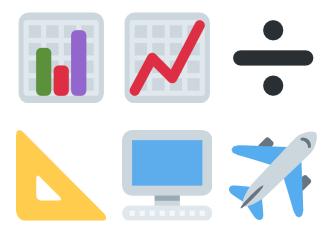
Python for Operational Research in Healthcare There's a library for that...

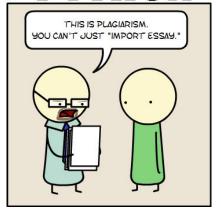
@GeraintPalmer

PyCon UK 2017





PYTHON



http://i.imgur.com/ZyeCO.jpg

The Basics

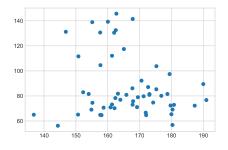
	Sex	Height	Weight
0	М	187.306088	72.233276
1	Μ	170.595112	92.195728
2	F	157.637346	64.835601
3	Μ	162.010640	130.462244
4	F	154.017198	81.568846
:	:	:	:

The Basics

	Sex	Height	Weight
0	М	187.306088	72.233276
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3	Μ	162.010640	130.462244
4	F	154.017198	81.568846
:	:	÷	÷

```
>>> import scipy.stats
>>> scipy.stats.ttest_ind(
...     df[df['Sex']=='M']['Height'],
...     df[df['Sex']=='F']['Height']
... ).pvalue
0.070033630470421021
```

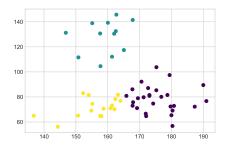
Machine Learning



```
>>> kmeans = KMeans(n_clusters=3, random_state=0)
>>> kmeans.fit(df[['Height', 'Weight']])
>>> df['Cluster'] = kmeans.labels_
```

>>> from sklearn.cluster import KMeans

Machine Learning



```
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```

>>> from sklearn.cluster import KMeans

Optimisation

7 8 9 9 7 5 4 8 4 3

























Full Time

 $\pounds 7.50$ per hour



4hrs, 1hr break, 3hrs



Part Time

£8 per hour



4hrs

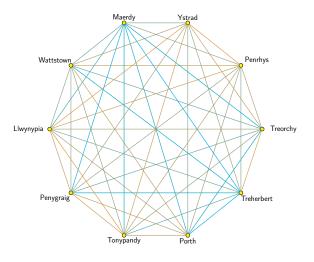
$$\underline{x} = [F_9, F_{10}, F_{11}, P_9, P_{10}, P_{11}, P_{12}, P_1, P_2, P_3]$$

C = [£60, £60, £60, £32, £32, £32, £32, £32, £32, £32]

```
>>> import pulp
>>> prob = pulp.LpProblem("Nurse Rostering", pulp.LpMinimize)
>>> x = pulp.LpVariable.dicts("x", range(10), cat=pulp.LpInteger)
>>> objective_funtion = sum(C[i] * x[i] for i in range(10))
>>> prob += objective_funtion
>>> for j in range(10):
... prob += sum(A[j][i] * x[i] for i in range(10)) >= B[j]
>>> for j in range(10):
\dots prob += x[j] >= 0
>>> prob.solve()
>>> [pulp.value(i) for i in x]
[-0.0, 1.0, 2.0, 7.0, -0.0, -0.0, -0.0, 4.0, -0.0, 1.0]
```

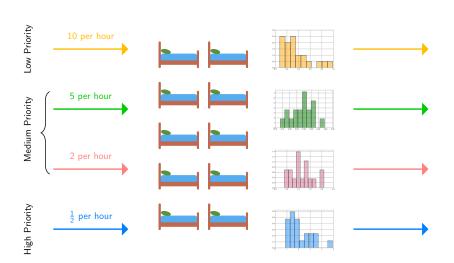
Graph Theory



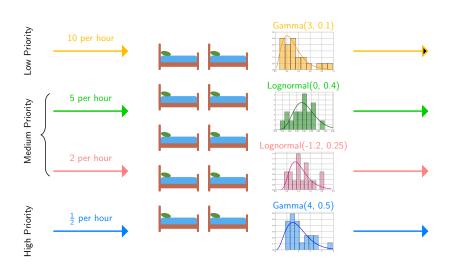


```
>>> import networkx as nx
>>> towns = ['Porth', 'Wattstown', 'Penrhys', 'Maerdy',
        'Penygraig', 'Tonypandy', 'Llwynypia', 'Ystrad'
        'Treorchy', 'Treherbert']
>>> D = nx.from_numpy_matrix(distances)
>>> D = nx.relabel nodes(D.
... {i: towns[i] for i in range(10)})
>>> ranks = nx.betweenness_centrality(D, weight='weight')
>>> max(ranks.keys(), key=lambda x: centrality[x])
'Ystrad'
```

Simulation

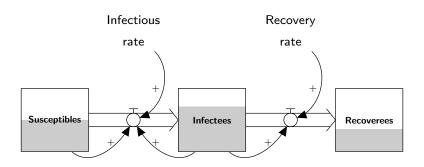


Simulation

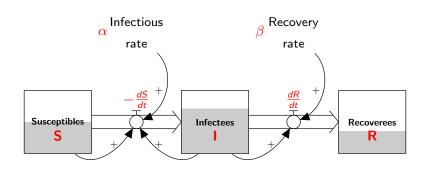


```
>>> import ciw
>>> N = ciw.create network(
        Arrival_distributions={
            'Class 0': [['Exponential', 10.0]],
            'Class 1': [['Exponential', 5.0]],
            'Class 2': [['Exponential', 2.0]],
            'Class 3': [['Exponential', 0.5]]},
        Service distributions={
            'Class 0': [['Gamma', 3, 0.1]].
            'Class 1': [['Lognormal', -1.2, 0.25]],
            'Class 2': [['Lognormal', 0, 0.4]],
            'Class 3': [['Gamma', 4, 0.5]]}.
        Number_of_servers=[10],
        Priority_classes={
            'Class 0': 2, 'Class 1': 1, 'Class 2': 1, 'Class 3': 0
. . . )
>>> average_waits = []
>>> for i in range(10):
       ciw.seed(1)
        Q = ciw.Simulation(N)
        Q.simulate_until_max_time(30)
       recs = Q.get_all_records()
        waits = [r.waiting_time for r in recs if r.arrival_date >= 6]
        average_waits.append(np.mean(waits))
>>> np.mean(average waits)
0.11887656711933872
```

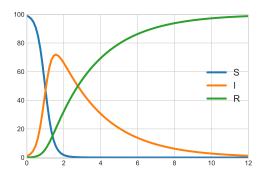
System Dynamics



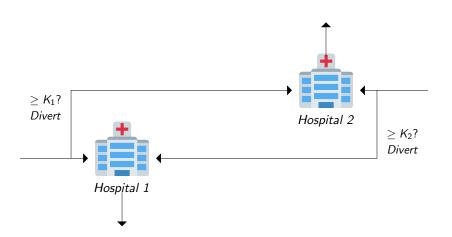
System Dynamics



$$\frac{\mathrm{d}S}{\mathrm{d}t} = -\alpha SI \qquad \qquad \frac{\mathrm{d}R}{\mathrm{d}t} = \beta I \qquad \qquad \frac{\mathrm{d}I}{\mathrm{d}t} = -\frac{\mathrm{d}S}{\mathrm{d}t} - \frac{\mathrm{d}R}{\mathrm{d}t}$$



Game Theory



$$K_2 = 0$$
 $K_2 = 40$ $K_2 = 65$ $K_2 = \infty$ $K_1 = 0$ $(59.6, 59.6)$ $(60.2, 59.6)$ $(51.8, 68.6)$ $(0.0, 119.6)$

$$K_1 = 0$$
 (59.6,59.6) (60.2,59.6) (51.8,68.6) (0.0,119.6) (59.6,60.2) (59.6,59.6) (50.6,67.9) (38.4,80.9)

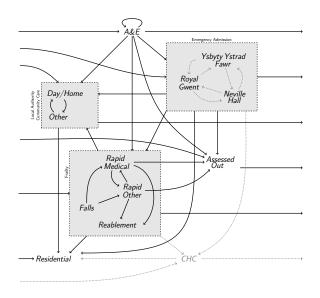
$$\begin{array}{lll} \mathcal{K}_1 = 0 \\ \mathcal{K}_1 = 40 \\ \mathcal{K}_1 = 65 \end{array} \left(\begin{array}{lll} (59.6, 59.6) & (60.2, 59.6) & (51.8, 68.6) & (0.0, 119.0) \\ (59.6, 60.2) & (59.6, 59.6) & (50.6, 67.9) & (38.4, 80.9) \\ (68.6, 51.8) & (67.9, 50.6) & (59.6, 59.6) & (56.7, 62.5) \end{array} \right)$$

$$K_1 = 40$$
 (59.6, 60.2) (59.6, 59.6) (50.6, 67.9) (38.4, 80.9)
 $K_1 = 65$ (68.6, 51.8) (67.9, 50.6) (59.6, 59.6) (56.7, 62.5)
 $K_1 = \infty$ (119.0, 0.0) (80.9, 38.4) (62.5, 56.7) (59.6, 59.6)

```
>>> g = nash.Game(-01, -02)
>>> eqs = g.support_enumeration()
>>> list(eqs)
[(array([ 0.,  1.,  0.,  0.]), array([ 0.,  1.,  0.,  0.]))]
```

>>> import nash

My PhD Work...



```
pip install pandas
pip install numpy
pip install matplotlib
pip install scipy==0.19.1
pip install sklearn==0.17.1
pip install pulp==1.6.8
pip install networkx==1.11
pip install ciw==1.1.3
pip install nashpy==0.0.11
```

Emojis thanks to Twemoji.