

Caminhadas aleatórias (Random Walks)

O primeiro projeto de baseia na implementação do cálculo da distribuição estacionária da cadeia de Markov a partir da matriz de transição de estados do jogo “Snakes and Ladders” (imagem1).

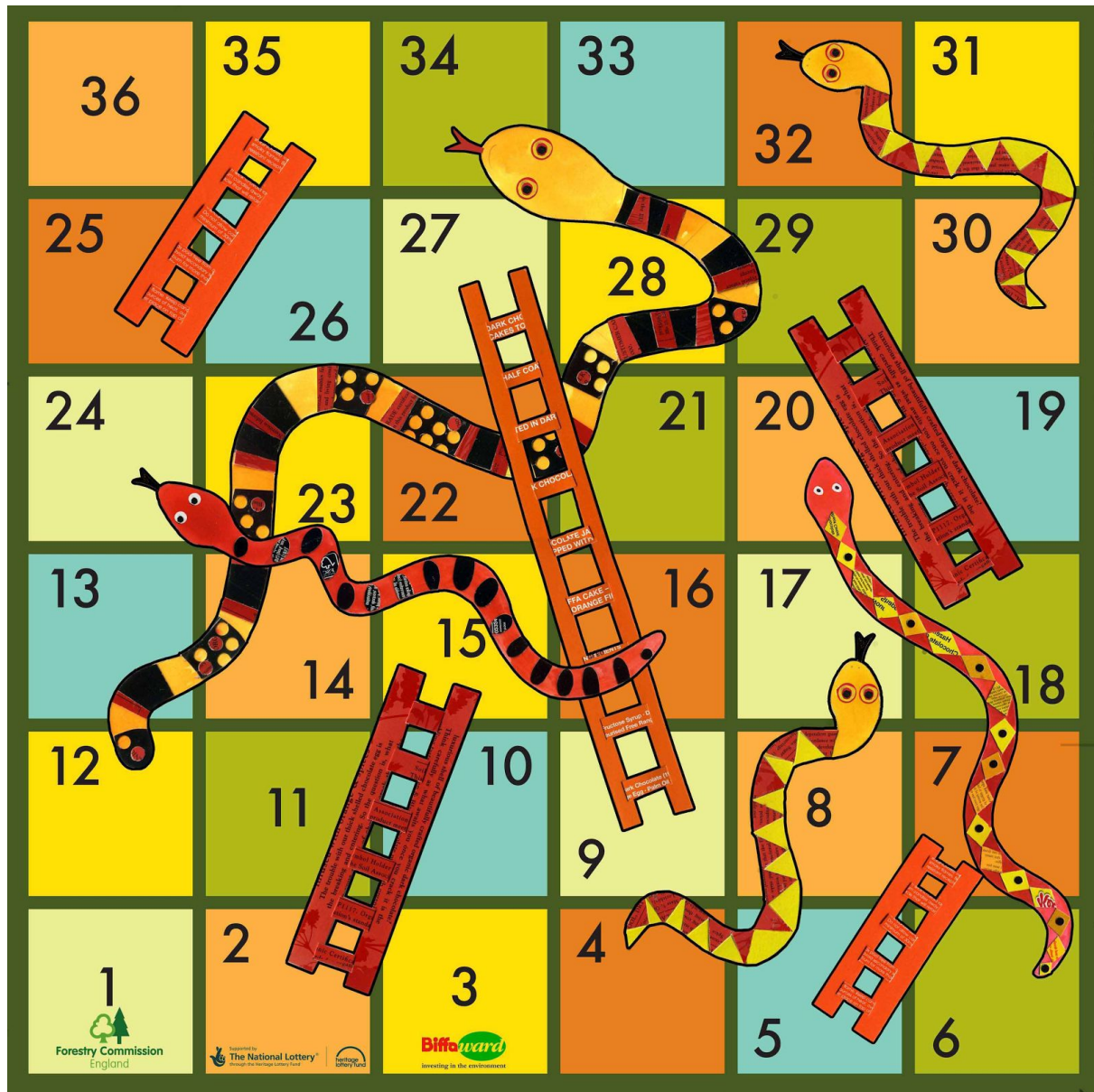


Imagem: Tabuleiro do jogo snakes and ladders.

Primeiramente foi elaborada a matriz de adjacências do tabuleiro representando o grafo do diagrama de estados presentes no jogo, a partir disso foi elaborada a matriz $(\Delta)^{-1}$, matriz essa que receberá em sua diagonal o valor $1/d(v_i)$, sendo $d(v_i)$ o grau dos vértices pertencentes ao grafo.

```

13 matrizADJ = [[0,1,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #1
14 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,0,0,0,0,0,0,0,0,0,0], #2
15 [0,0,0,1,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #3
16 [0,0,0,0,1,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #4
17 [0,0,0,0,0,0,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #5
18 [0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #6
19 [0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #7
20 [0,0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #8
21 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,0,0], #9
22 [0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0,0,0,0,0], #10
23 [0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0,0], #11
24 [0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0,0], #12
25 [0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0,0], #13
26 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0], #14
27 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0,0], #15
28 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0,0,0,0,0], #16
29 [0,0,0,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #17
30 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,0,0], #18
31 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0], #19
32 [0,0,0,0,0,1,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], #20
33 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0], #21
34 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0], #22
35 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0,0,0,0], #23
36 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,0,0,0,0,0,0,0,0,0], #24
37 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1], #25
38 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0], #26
39 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0], #27
40 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0], #28
41 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0], #29
42 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0], #30
43 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0], #31
44 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,0,0], #32
45 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1,1,0], #33
46 [0,0,0,0,0,0,0,0,0,0,0,0,1,0,0,0,0,0,0,0,0,0,0,0,0], #34
47 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1], #35
48 [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,1]] #36
```

Imagem: Matriz de adjacência.

A matriz P, ou matriz de probabilidade é computada como sendo: $P = (\Delta)^{-1} * \text{matrizAdj}$, a matriz de probabilidades define o processo conhecido como Cadeia de Markov.

```
matrizD = []
max_linhas = 36
max_colunas = 36
soma = 0
k = 0
```

```
#Computando matriz delta:
```

```
for i in range(max_linhas):
    linha = []
    for j in range(max_colunas):
        linha.append(0)
    matrizD.append(linha)
```

```

for i in range(max_linhas):
    for j in range(max_colunas):
        soma = soma + matrizADJ[i][j]
    matrizD[k][k] = 1/soma
    k = k+1
    soma = 0

```

#computando matriz P = matrizD*matrizADJ

```

matrizP = []
mult = 0

```

```

for i in range(max_linhas):
    linha = []
    for j in range(max_colunas):
        for k in range(max_colunas):
            mult = mult + matrizD[i][k]*matrizADJ[k][j]
        linha.append(mult)
        mult = 0
    matrizP.append(linha)

```

Para computar o vetor de probabilidades de cada estado num tempo k, nesse exercício k = 100, 'w' da cadeia de Markov Homogênea, foi utilizado o Power Method, que consiste de multiplicar a matriz de probabilidades por ela mesma k vezes, e depois obter a partir do resultado um vetor contendo a probabilidade de se estar em cada estado em um tempo k.

```

# tempo k
for k in range(100):
    aux = []
    #multiplicação
    for i in range(36):
        aux.append(0)
        for j in range(36):
            aux[i] += v[j]*matrizP[j][i]
    v = aux

```

Resultado Obtidos

```

[0.0,
0.0,
0.0,

```

0.0019876139002673876,
 0.0010245929380067287,
 0.0010245929380067287,
 0.0015844982582033495,
 0.001344957181767505,
 0.001510101754845458,
 0.0006933106591457362,
 0.00035739421106157796,
 0.0024530652647714985,
 0.0014487609716806375,
 0.002011348331737679,
 0.0017836480208753363,
 0.0019562787784050436,
 0.001927891325030704,
 0.0010084403567855687,
 0.0,
 0.0,
 0.0,
 0.0,
 0.0,
 0.0,
 0.0,
 0.0,
 0.0015568823432814804,
 0.0008025561497204872,
 0.0022559449592002463,
 0.011278882947334912,
 0.006977054299492151,
 0.009410734123282851,
 0.003596594392709062,
 0.0018540048143700673,
 0.0018540048143700673,
 0.9382968462656476]

A matriz P Barra referente ao Page Rank foi obtida através da expressão:

$P = (1 - \alpha) P + \alpha \frac{1}{n} U$ (Google matrix) , onde
 $\alpha = 0.1$

P = matriz de Probabilidades

U = matriz composta unicamente por 1's

Para facilitar a o cálculo da google Matrix, utilizamos recursos presentes na biblioteca networkX, outro grafo sendo equivalente ao primeiro foi gerado, dessa vez com a criação do grafo utilizando da biblioteca e inserção dos vértices e arestas:

```
vertices = range(1, 37)
arestas = [(1, 2), (1, 3), (2, 15), (3, 4), (3, 5), (4, 5), (4, 6), (5, 7), (6, 7), (6, 8), (7, 8), (7, 9), (8, 9), (8, 10), (9, 27),
            (10, 11), (10, 12), (11, 12), (11, 13), (12, 13), (12, 14), (13, 14), (13, 15), (14, 15), (14, 16), (15, 16), (15, 17), (16, 17),
            (16, 18), (17, 4), (18, 29), (19, 20), (19, 21), (20, 6), (21, 22), (21, 23), (22, 23), (22, 24), (23, 24), (23, 25), (24, 16),
            (25, 35), (26, 27), (26, 28), (27, 28), (27, 29), (28, 29), (28, 30), (29, 30), (29, 31), (30, 31), (30, 32), (31, 32),
            (31, 33), (32, 30), (33, 34), (33, 35), (34, 12), (35, 36), (36, 36)]
```

```
grafo = nx.DiGraph(name = 'Dígrafo snakes and ladders')
```

```
grafo.add_nodes_from(vertices)
grafo.add_edges_from(arestas)
```

Foi criada a matriz U a calculada o Page rank através das seguintes instruções, em seguida utilizando também do Power Method, foram computados os dois resultados obtidos.

```
#matriz P_
mP_ = matrizP[i][j]*(1-alfa) + u * (alfa * (1/grafo.number_of_nodes()))

#POWER METHOD COM MATRIZ P BARRA
mP_aux = mP_
#P elevado a k
for _ in range(0, k-1):
    mP_aux = np.dot(mP_aux, mP_)

#w(100) = w(0) * P^k
wk_ = np.dot(w, mP_aux)
```

Como observado nos resultados obtidos através do Power Method utilizando-se Pagerank e Cadeia de Markov, obtivemos resultados diferentes sendo o primeiro mais preciso pois com o Pagerank não são obtidos resultados nulos na distribuição estacionária.

