

a) state the transition matrix P for MC(a)

```
import math
import numpy as np
```

```
#main
Pa = np.matrix(
(
(0,0.5,0.5,0,0,0),
(1,0,0,0,0,0),
(0,0.9,0,0.1,0,0),
(0,0,0,0,1,0),
(0,0,0,0,0,1),
(0,0,0,1,0,0)
)
)
```

b) irreducibility test over a MC defined by transition matrix P

```
import math
import numpy as np
import networkx as nx
```

```
def DFSrec(G,v,visited):
    visited[v]= True

    for i in list(G.neighbors(v)):
        if visited[i]==False:
            DFSrec(G, i, visited)

def DFS(G, startAt):
    visited = [False]*(G.number_of_nodes())
    DFSrec(G,startAt,visited)

    if any(i == False for i in visited):
        return False

    return True

def isStronglyConnected(G):
    if DFS(G,0) == False:
        return False

    # returns a graph with the same set of V and E of G but having all
    # directions inverted
    T = G.reverse(True)
    return DFS(T,0)

def createGraphFromMC(P):
    # Create Directed Graph
    G=nx.DiGraph()

    # Add a list of nodes:
    G.add_nodes_from(range(0,len(P)))

    # Add a list of edges:
```

```

    for i in range(0,len(P)):
        for j in range(0,len(P)):
            if P[i,j]>0:
                G.add_edge(i,j)

    return G

def isIrreducibleMC(P):
    G = createGraphFromMC(P)
    print "irreducible" if isStronglyConnected(G) else "NOT irreducible"

```

thinking of the MC as a directed graph $G(V,E)$, the MC is irreducible if and only if the G is strongly connected. This condition can be assessed in many ways. A longer one consists in applying a DepthFirstVisit of G for all its vertices $v \in V$. If for each $v \in V$ the DFS is able to visit other vertices, G is strongly connected.

A shorter approach (Kosaraju) provides an answer applying only two DFS at the same v , the second of which is run after having inverted the edges of G .

c) periodicity test over a MC defined by transition matrix P (to be adjusted possibly)

d) power of transition matrix

```

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(0,0,0,0,0,1),
(0,0,0,1,0,0)
)
)

```

```

Pb = np.matrix(
(
(0,0.5,0.5,0,0,0),
(1,0,0,0,0,0),
(0,0.9,0,0.1,0,0),
(0,0,0,0,1,0),
(0,0,0,0,0,1),
(0.5,0,0,0.5,0,0)
)
)

```

```

print Pa**50

```

```
print Pb**50
```

e) invariant distribution π over a MC defined by transition matrix P

```
import math
import numpy as np

# https://stephens999.github.io/fiveMinuteStats/stationary_distribution.html
def findInvDistrib(P):

    I = np.identity(len(P))

    # add condition over pi all entries sum to 1
    newrow = np.ones(len(P))
    A = P
    A = A.transpose()
    A = np.vstack([A-I, newrow])
    print A

    B = np.zeros(len(P)+1)
    B[len(P)] = 1

    x, residuals, rank, s = np.linalg.lstsq(A, B)
    print "Solution: ", x
    print "Residuals: ", residuals
    print "Rank: ", rank
    print "Test1: ", x*P
    print "Test2: ", P**1000

'''
    try:
        x = np.linalg.solve(A, B)
        print 'Invariant distribution found'
        print x
    except np.linalg.LinAlgError as err:
        if 'Singular matrix' in str(err):
            print 'Singular matrix passed'
        else:
            raise
'''

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(0,0,0,1,0),
(0,0,0,0,1),
(0.5,0,0,0.5,0,0)
)
)
```

```
findInvDistrib(Pa)
findInvDistrib(Pb)
```

A system of equations can be set to solve $\pi P = \pi$ adding the following constraint over π :

$$\pi_1 + \pi_2 + \dots + \pi_n = 1$$

Since the matrix defining the system is no longer square Python `linalg.solve` cannot be applied. We can use `linalg.lstsq` anyway . If residuals are vanishingly small we can assume the solution is unique and compare the result with conditions over the MC required for uniqueness of π to hold (any irreducible and aperiodic MC on a finite state space as a unique steady distribution).