

project 15: more on Markov chains**task 15.1: products of column stochastic matrices and vectors**

Recall that a column stochastic matrix is a square matrix $P \in \mathbb{R}^{m \times m}$ where

$$P_{ji} \geq 0 \quad \text{and} \quad \sum_{j=1}^m P_{ji} = 1$$

and that an m -dimensional stochastic vector is a vector $u \in \mathbb{R}^m$ where

$$u_i \geq 0 \quad \text{and} \quad \sum_{i=1}^m u_i = 1$$

Now, given P and u as above, prove that the vector

$$v = P u$$

is a stochastic vector, too.

task 15.2: products of column stochastic matrices

Prove that, if $P, Q \in \mathbb{R}^{m \times m}$ are column stochastic matrices, then

$$R = P Q$$

is a column stochastic matrix, too.

task 15.3: estimating Markov chains

Load the 3D data points in the file `q3dm1-path1.csv` into a data matrix X . Using python's `numpy` module, this can be accomplished like so

```
import numpy as np
matX = np.loadtxt('q3dm1-path1.csv', delimiter=',')
matX = matX.T
```

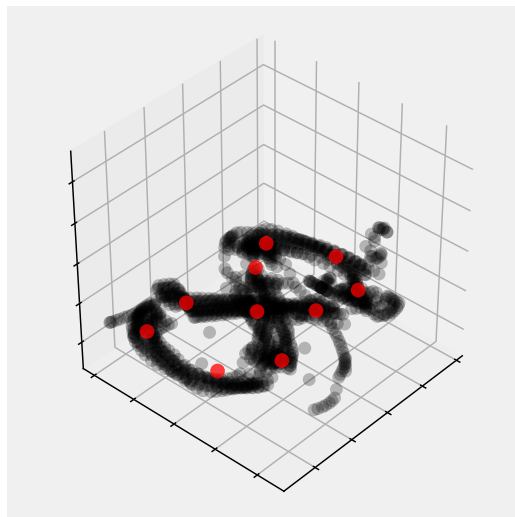
Note: if you know what you are doing (in the next couple of computational steps), then transposition of the array `matX` is not really necessary. However, as your instructor likes his matrices to be column matrices, all the following code examples assume `matX` has been transposed.

Now, cluster the 1288 columns of X into $k = 10$ clusters. If you do this using

```
import scipy.cluster.vq as vq
matM, inds = vq.kmeans2(matX.T, k=10, iter=100, minit='++')
matM = matM.T
```

you will obtain a matrix $M \in \mathbb{R}^{3 \times 10}$ whose columns m_i contain the cluster prototypes.

Plot the data in X and the prototypes in M . Your result should look something like this



If it does not, then maybe you need to run `kmeans2` several times until your result is satisfactory.

Now, recall that the 3D data points in `q3dm1-path1.csv` resulted from recording movements of a human player on the Quake III map `q3dm1`. In other words, we can understand the columns x_t of matrix X as the elements $x[t]$ of a sequence of locations on that map.

Here is a crazy idea: We can also understand the columns m_i of matrix M as 3D representations of the states of a discrete state space

$$S = \{s_1, s_2, \dots, s_k\}$$

such that

$$\begin{aligned} s_1 &\Leftrightarrow m_1 \\ s_2 &\Leftrightarrow m_2 \\ &\vdots \\ s_k &\Leftrightarrow m_k \end{aligned}$$

If so, then we can turn the sequence

$$X = x[1] : x[2] : x[3] : \dots \in \mathbb{R}^{3*}$$

into a sequence

$$s = s[1] : s[2] : s[3] : \dots \in S^*$$

where

$$s[t] = s_b$$

and the index of *best matching unit* is

$$b = \operatorname{argmin}_i \|x[t] - m_i\|^2$$

Implement this idea! That is, write code that takes the columns $x[t]$ of X and produces a sequence s of states $s[t] \in S$.

Next, given s , estimate a Markov chain λ that most likely has produced this sequence. That basically is, estimate the corresponding matrix $P_\lambda \in \mathbb{R}^{k \times k}$

of state transition probabilities. Recall that maximum likelihood estimates of the entries of P_λ are given by

$$(P_\lambda)_{ji} = \frac{n_{ji}}{\sum_{j=1}^k n_{ji}}$$

where n_{ji} counts the number of transitions $s_i \rightarrow s_j$ in sequence s .

If you implement this procedure using `numpy`, you will obtain an array, say, `matP`. Print this array using

```
print (np.round(matP, 2))
```

Discuss what do you observe? Your result should be a matrix of a (more or less) special structure ...

Finally, repeat all of this for different choices of k . Also, it might be a good idea to warm up to the data in file `q3dm1-path2.csv`. So, do all of the above once again but for the data in `q3dm1-path2.csv`.