

MARKOV CHAINS

UNDERSTANDING HOW THEY WORK

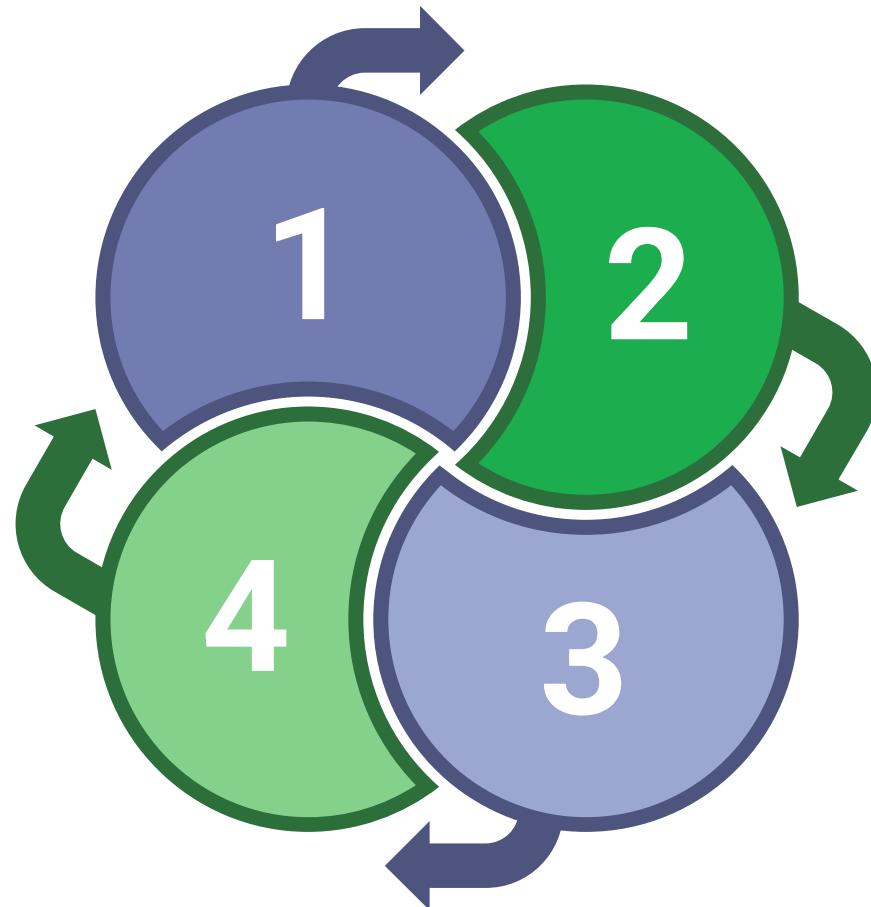
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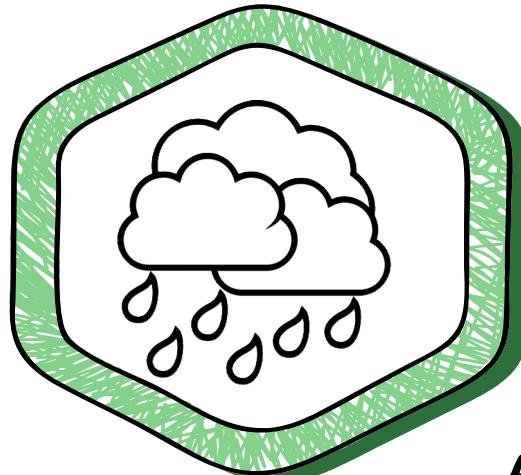




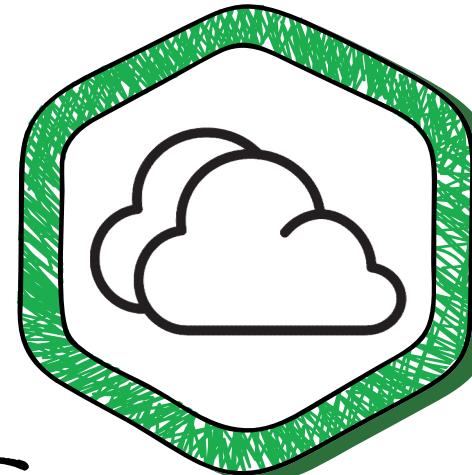
I. A brief overview

Weather states

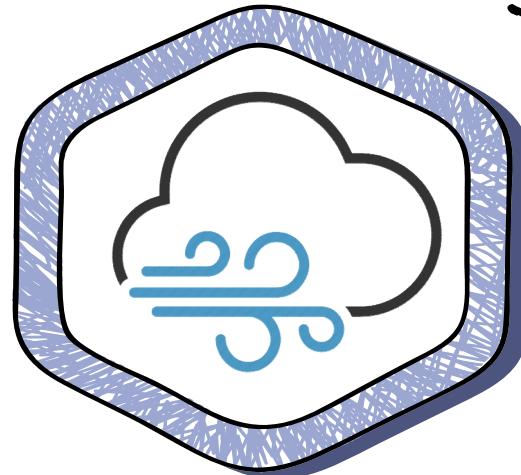
Rainy



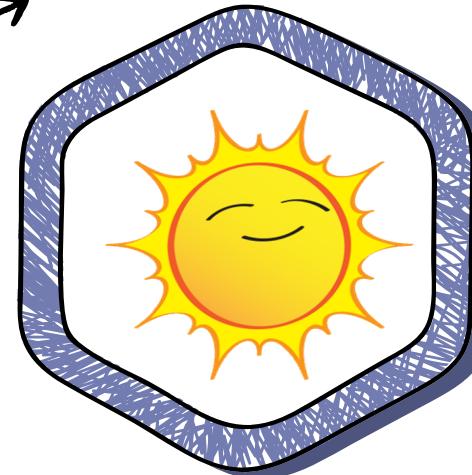
Cloudy



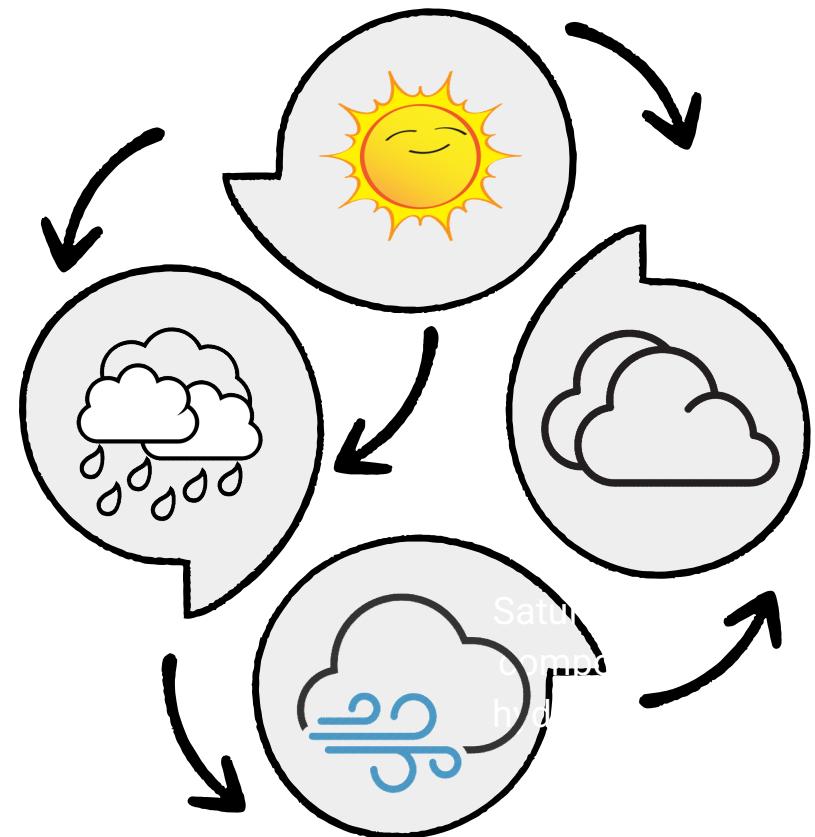
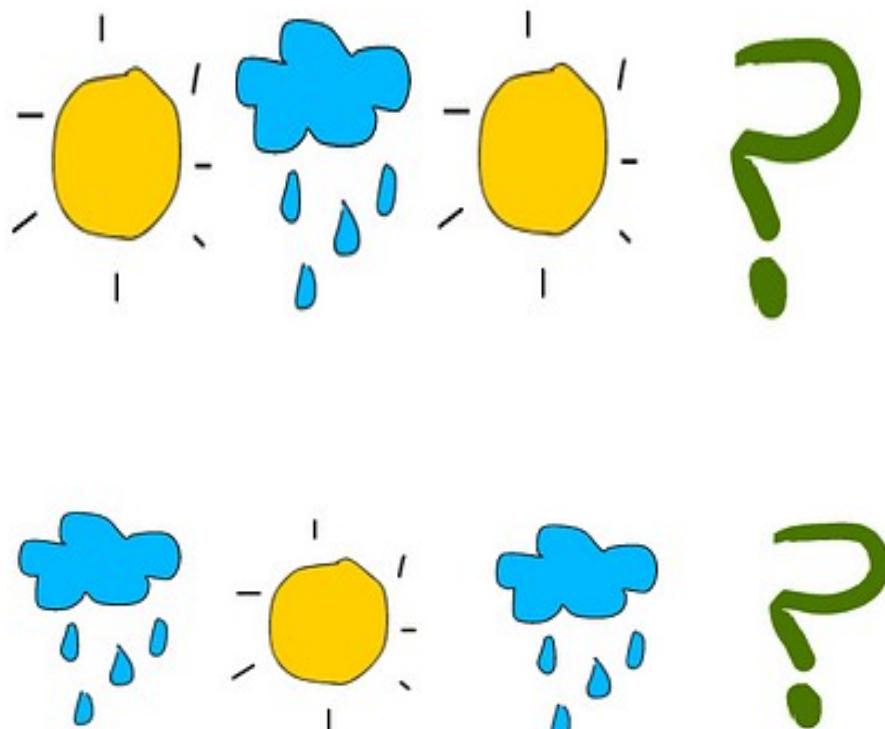
Windy



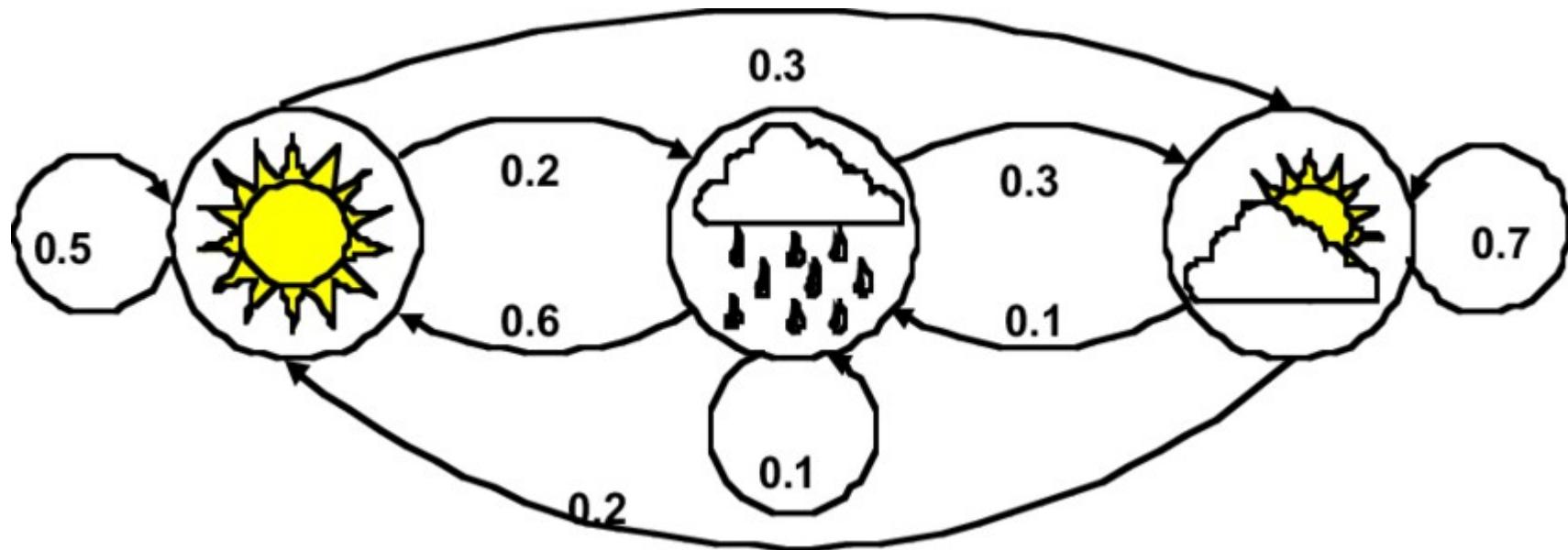
Sunny



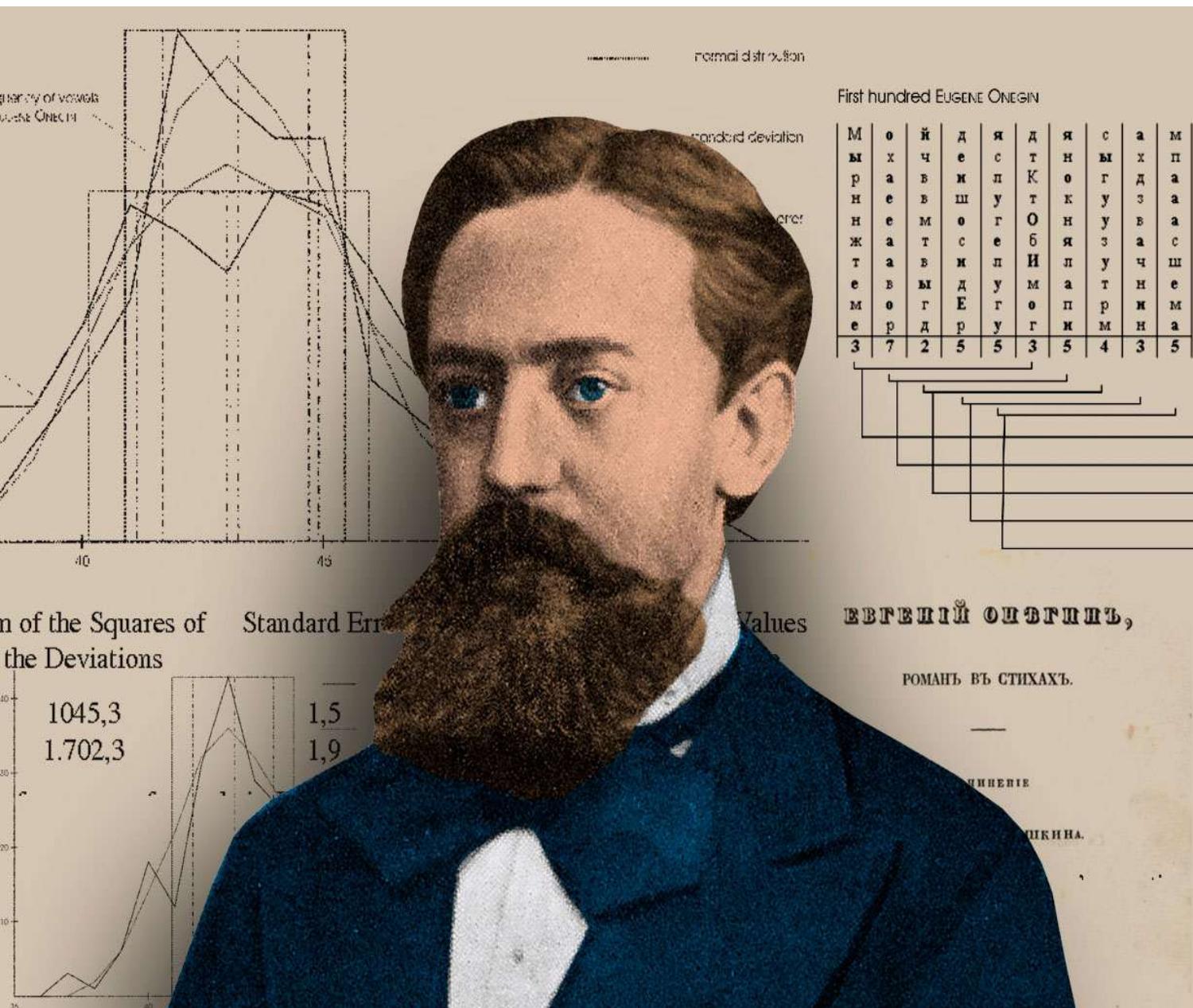
Weather forecasting? How it works?



Weather forecasting? How it works?



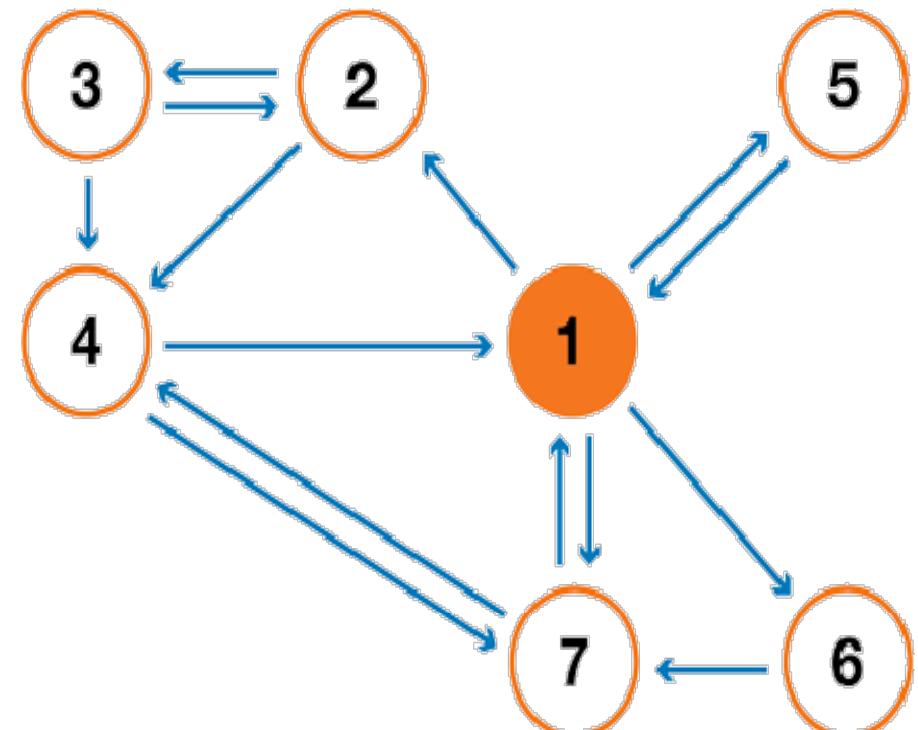
Everything existing in the universe is the fruit of **chance**.
-- Democritus



II. Formal definition

Definition

A **Markov Chain** is a stochastic model describing a sequence of possible events in which the probability of each event depends only on the state attained in the previous event.

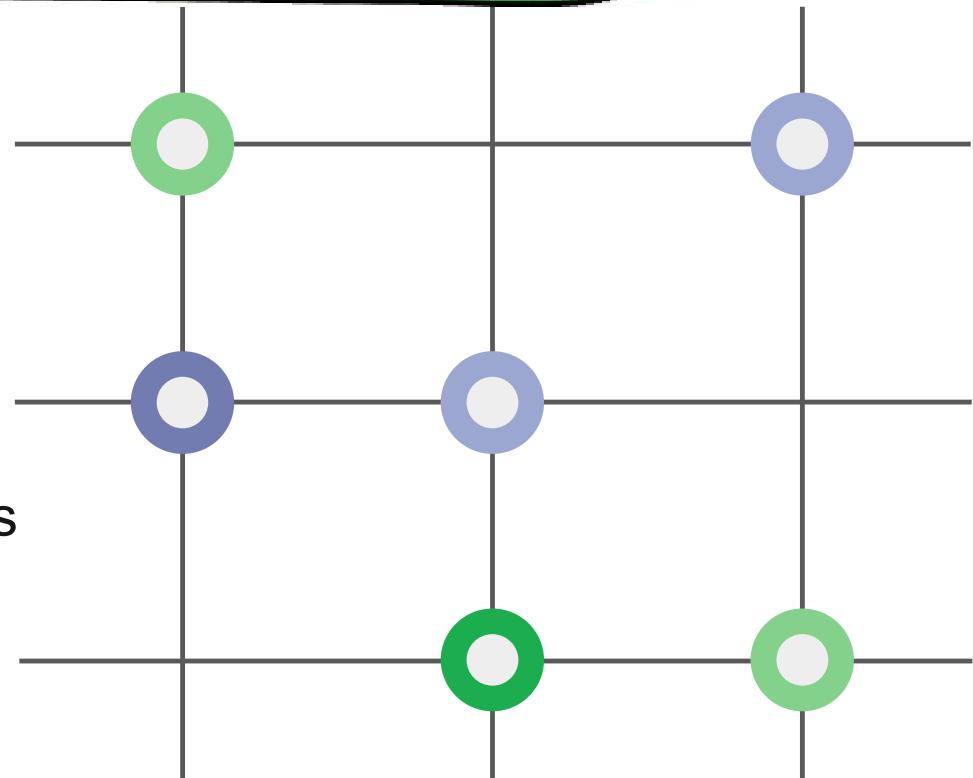


Trajectory : 1

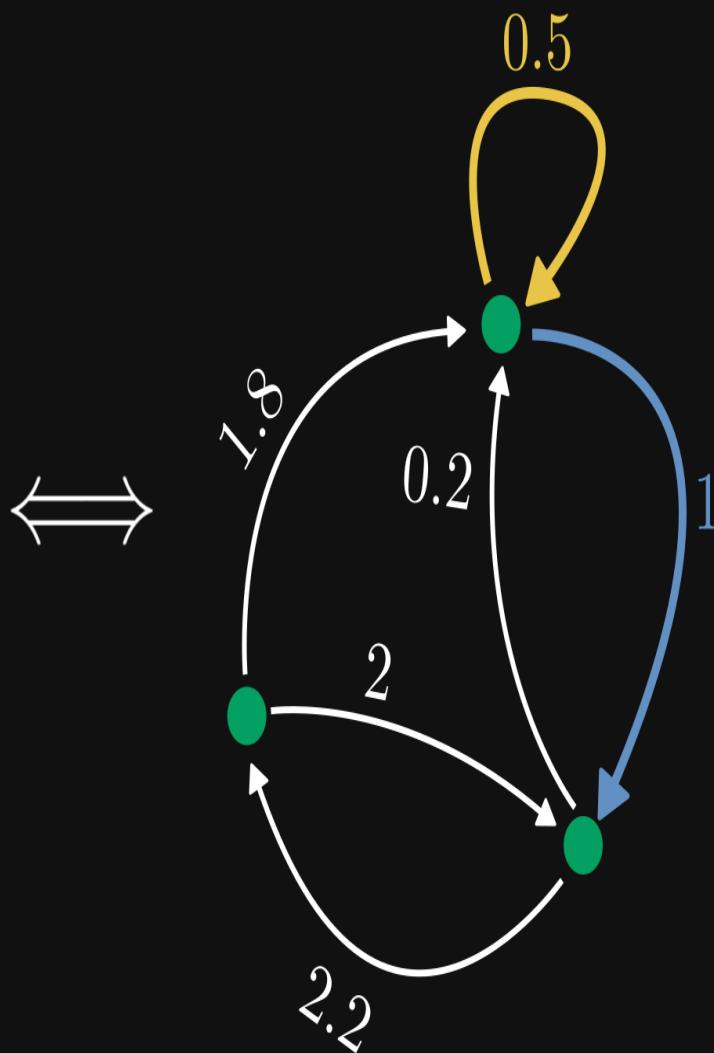
The Markov Property: “memorylessness”

$$P(X_n=i_n | X_0=i_0, X_1=i_1, \dots, X_{n-1}=i_{n-1}) = P(X_n=i_n | X_{n-1}=i_{n-1})$$

- For any positive **integer n** and possible states i_0, i_1, \dots, i_n of the random variables
- Knowledge of the **previous state** is all that is necessary to determine the probability distribution of the **current state**.



$$\begin{bmatrix} 0.5 & 1 & 0 \\ 0.2 & 0 & 2.2 \\ 1.8 & 2 & 0 \end{bmatrix}$$



III. Transition matrices

Transition Matrices

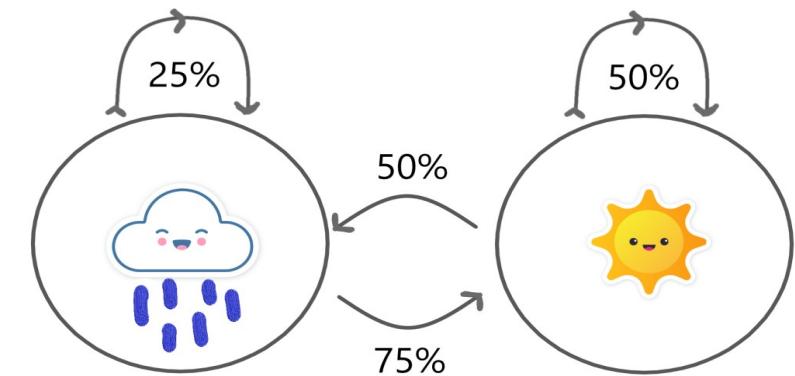
$$(P_t)_{i,j} = P(X_{t+1}=j \mid X_t=i).$$

- Contain information on the **probability of transitioning** between states.
- Each row of the matrix is a **probability vector**, and the sum of its entries is **1**.

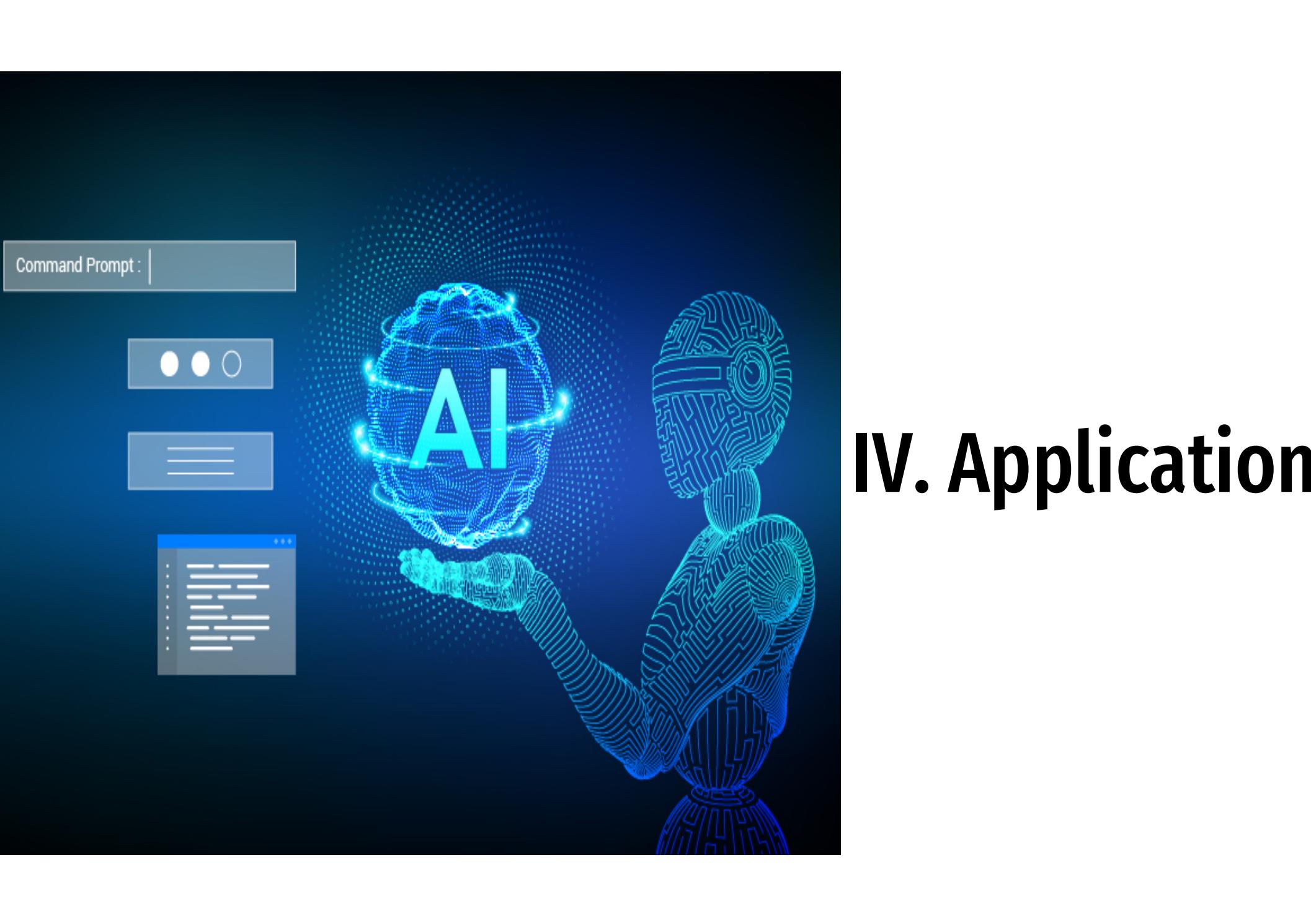
$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Transition matrix example

		Rainy	Sunny
Rainy	0.25	0.75	
Sunny	0.5	0.5	



[Visit: https://setosa.io/ev/markov-chains/](https://setosa.io/ev/markov-chains/)



IV. Applications

AI text generation

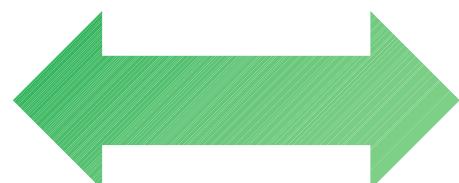
- Markov Chains can be used to **generate text** by **modeling the probability** of a word or character following another.
- By training the model on a large corpus of text, it can learn the probabilities of word or character sequences and **generate new text** that **resembles the training data**.



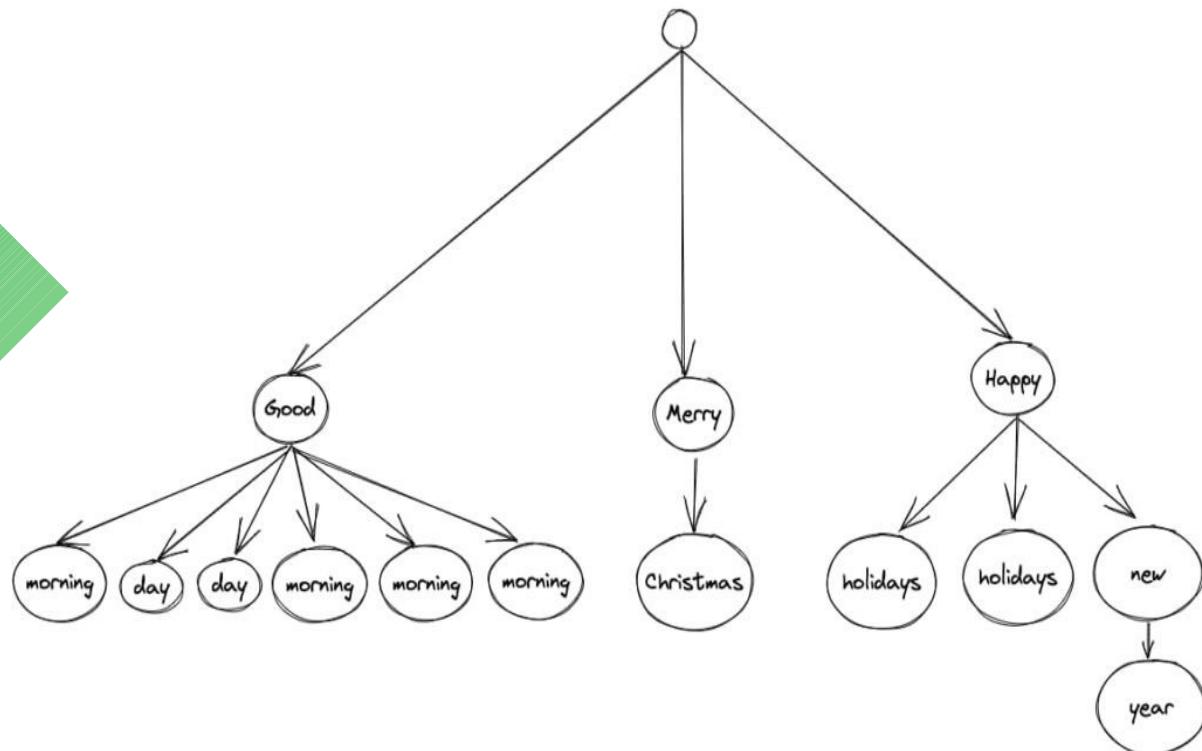
AI Text Generation

Imagine we have a list of standard greetings:

1. Good morning
2. Good day
3. Merry Christmas
4. Good morning
5. Good morning
6. Good day
7. Happy holidays
8. Happy New Year
9. Happy holidays
10. Good morning



We can **visualize** them by connecting each word together in a **graph**,



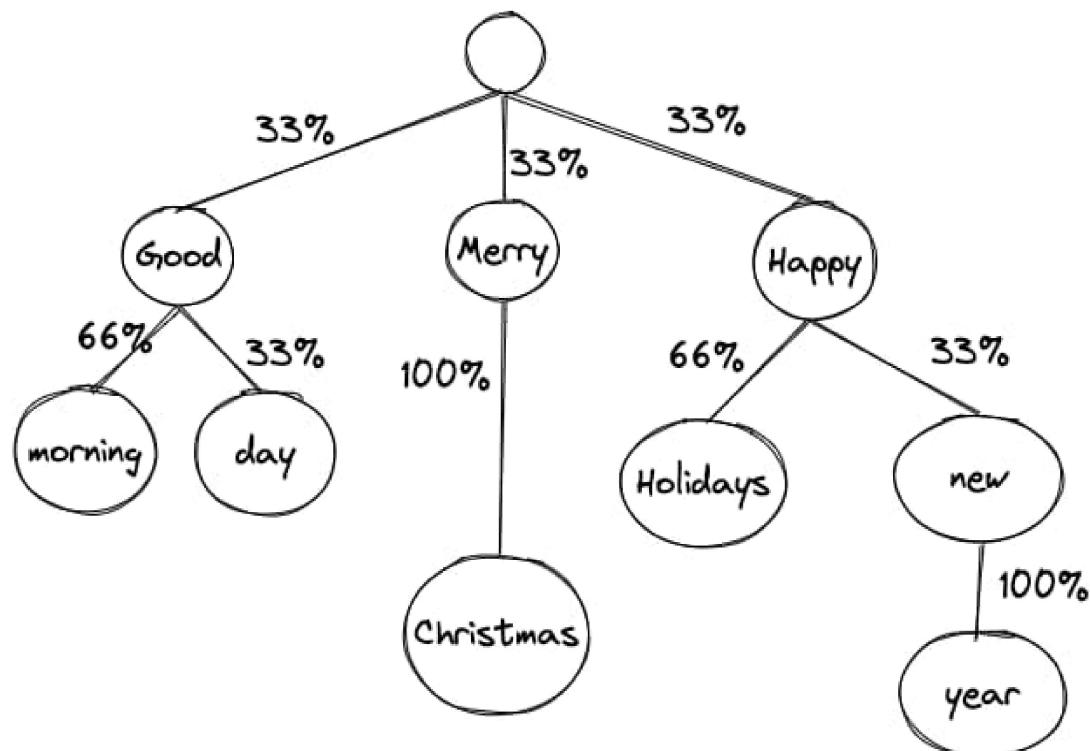
The creative possibilities

Going through that exercise we can make a few conclusions:

- There is a **1/3 chance** to start with “good”, “merry”, or “happy” respectively.
- If we say “good”, there is a **1/3 chance** we should say “day” and a **2/3 chance** we should say “morning”.
- If we say “merry”, we should say “Christmas”.
- If we say “happy”, there is a **2/3 chance** we should say “holidays” and a **1/3 chance** we should say “New Year”.



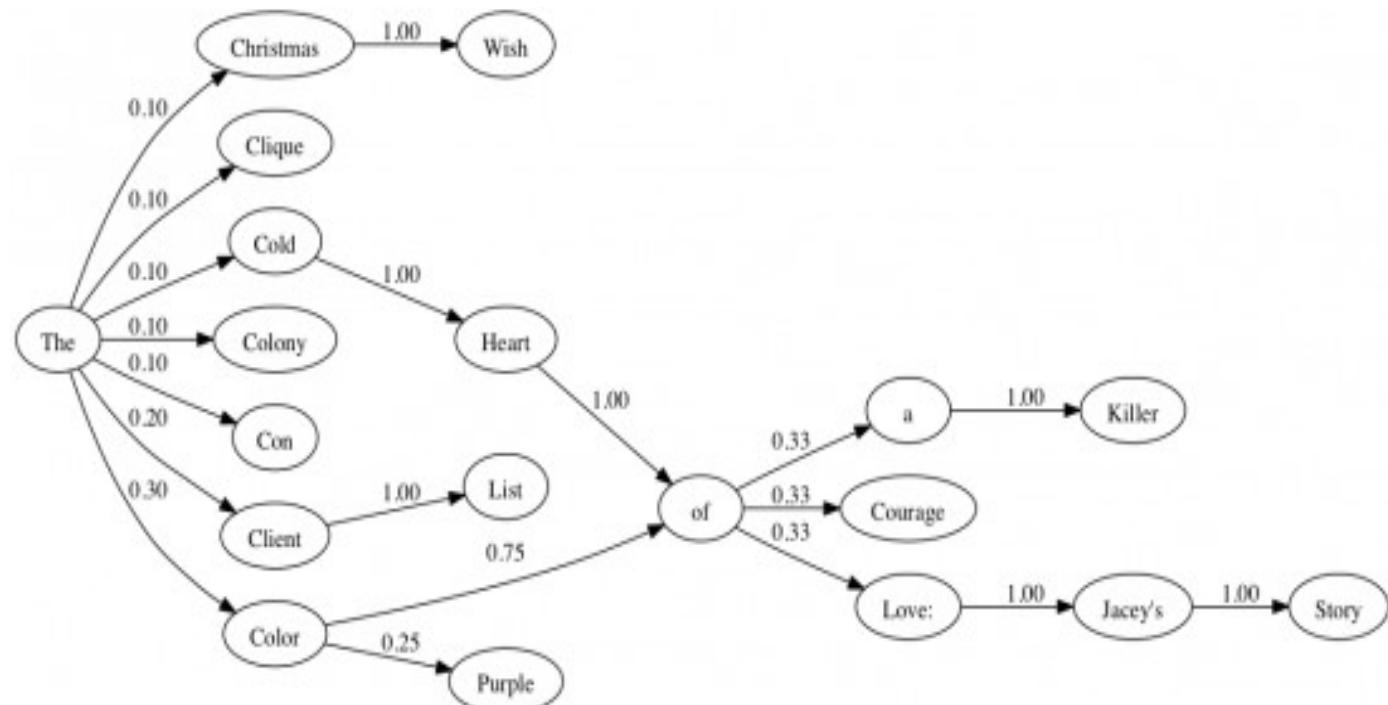
We redraw that graph to display those percentages, we get a new graph.



AI Text Generation

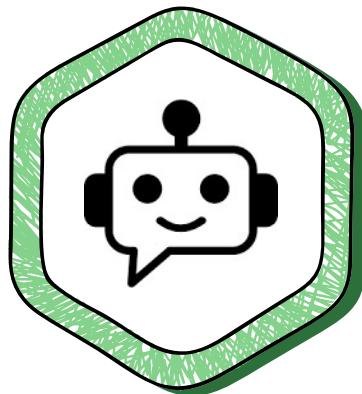
To build a **Markov chain algorithm** that can generate random text, we can follow these steps:

- Find a text **source**
- **Tokenize** the text
- Build the **Markov chain**
- **Traverse** the Markov chain

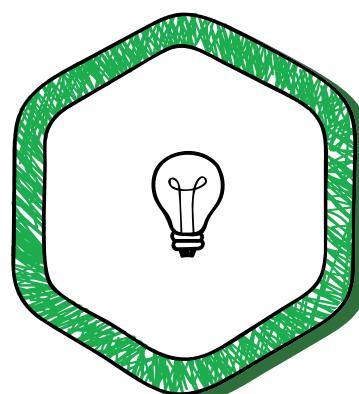


Visit: <https://projects.haykranen.nl/markov/demo/>

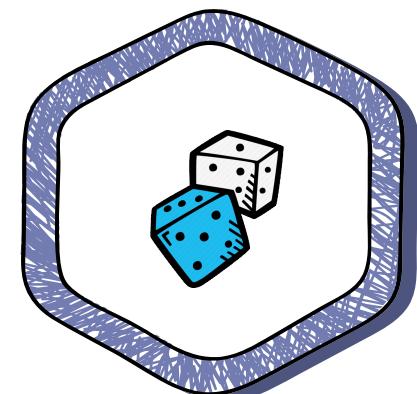
Common applications of Text Generation with Markov Chains



ChatBots

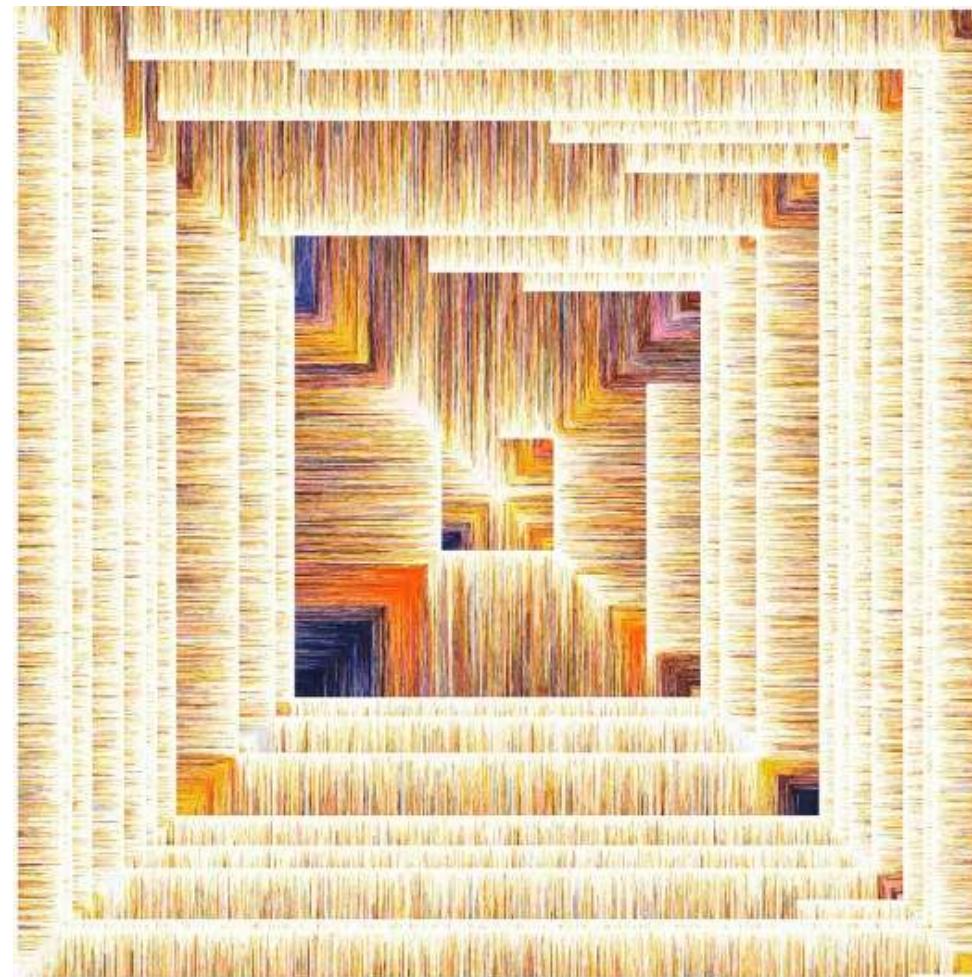
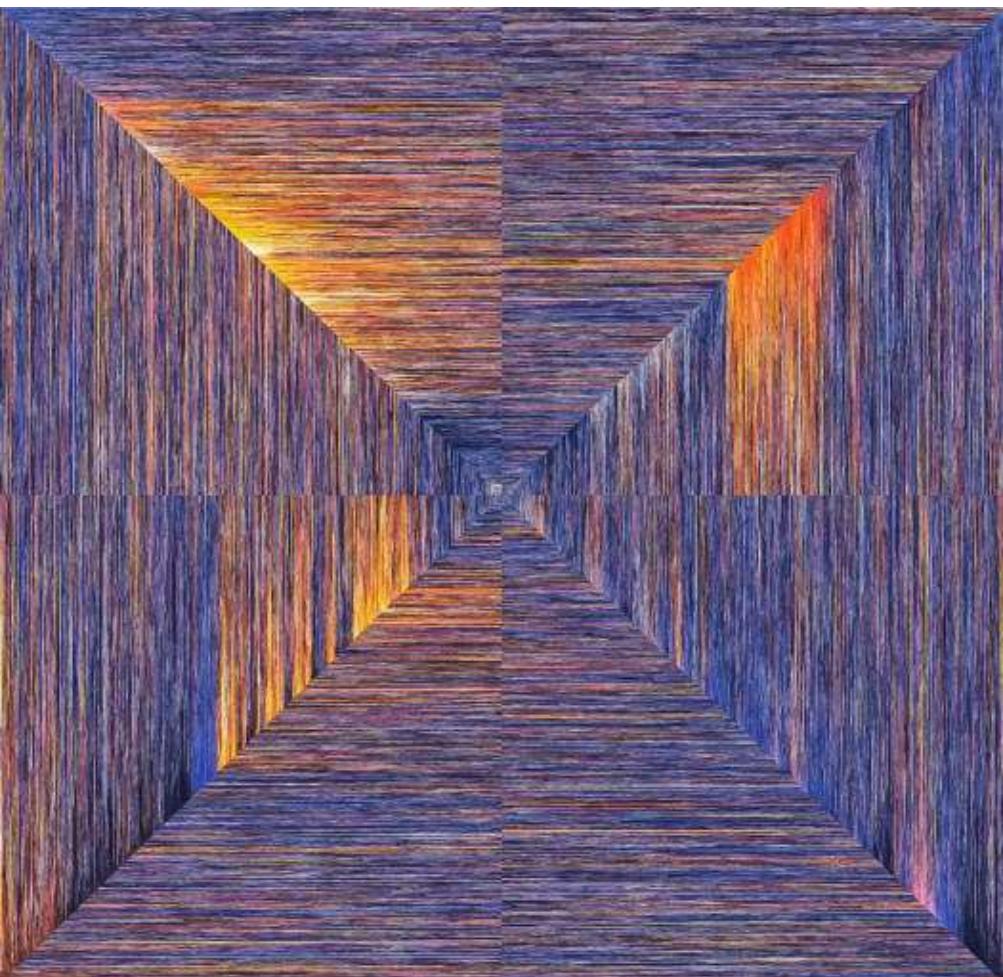


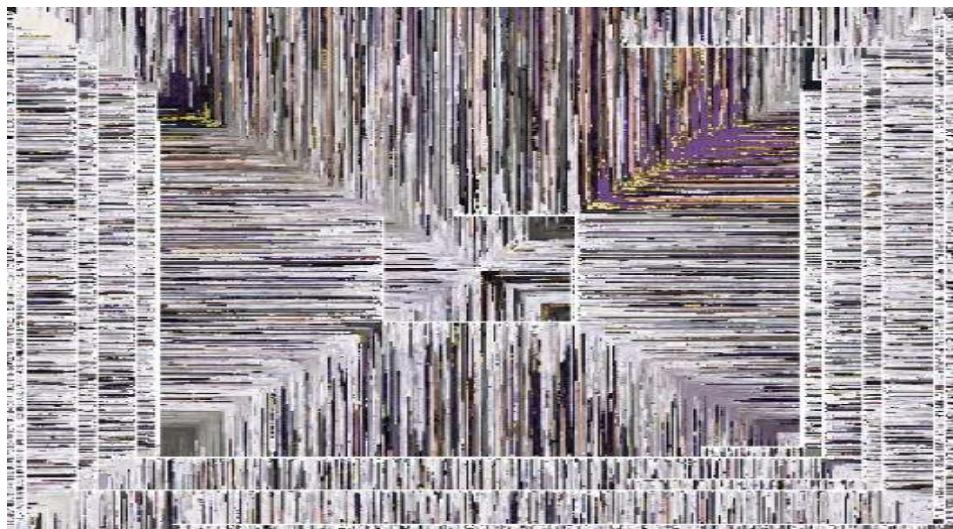
**Creative story
writing**



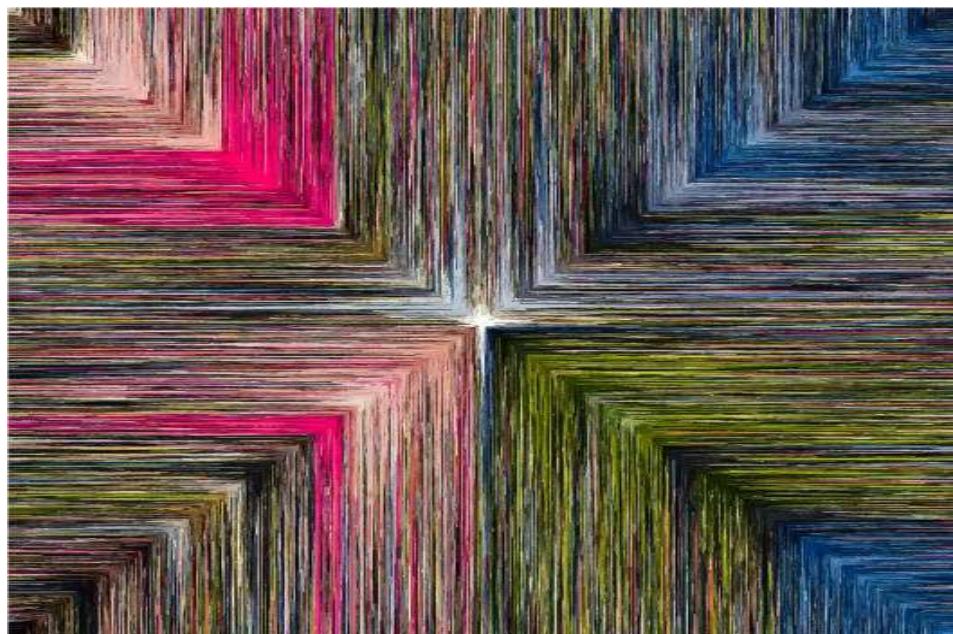
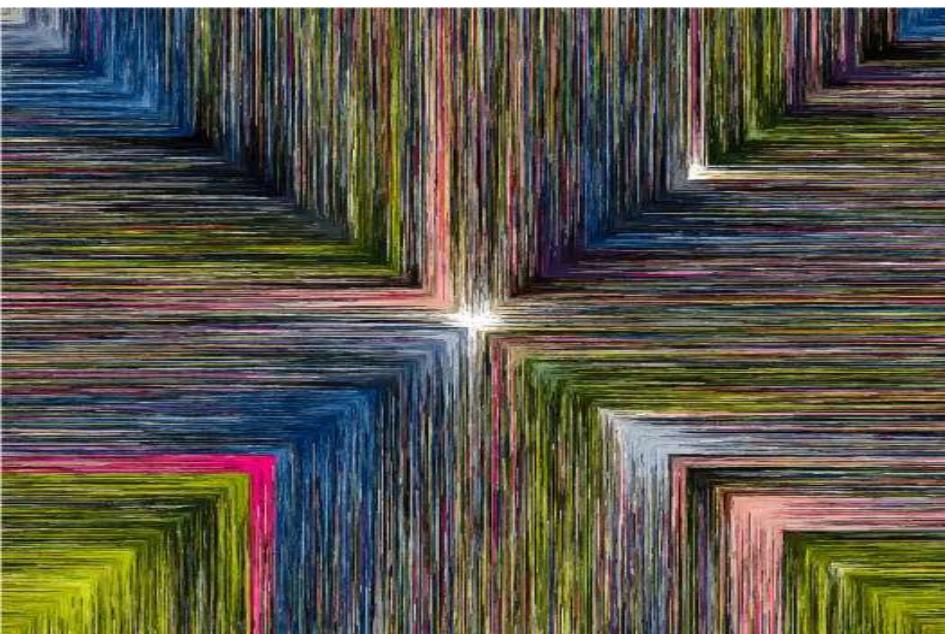
Just for fun !

Mathematical Art





Visit: <https://www.timswast.com/blog/2013/08/07/markov-chains-tutorial/>



Conclusion

