

DP IB Maths: AI HL



Your notes

4.13 Transition Matrices & Markov Chains

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4.13.1 Markov Chains

Markov Chains

What is meant by a “state”?

- States refer to **mutually exclusive events** with the current event **able to change over time**
- Examples of states include:
 - Daily weather conditions
 - The states could be: “sunny” and “not sunny”
 - Countries visited by an inspector each day
 - The states could be: “France”, “Spain” and “Germany”
 - Store chosen for weekly grocery shop:
 - The states could be: “Foods-U-Like”, “Smiley Shoppers” and “Better Buys”

What is a Markov chain?

- A **Markov chain** is a model that describes a **sequence of states** over a period of time
 - Time is measured in discrete steps
 - Such as days, months, years, etc
- The **conditions** for a Markov chain are:
 - The **probability** of a state being the **next state** in the sequence **only depends** on the **current state**
 - For example
 - The 11th state **only depends** on the 10th state
 - The first 9 states **do not affect** the 11th state
 - This probability is called a **transition probability**
 - The **transition probabilities do not change** over time
 - For example
 - The probability that the 11th state is A given that the 10th state is B is equal to the probability that the 12th state is A given that the 11th state is B
- A Markov chain is said to be **regular** if there is a value k such that in **exactly k steps** it is possible to reach any state regardless of the initial state
 - The chain where A can only go to B, B can only go to C and C can only go to A, is **not regular**
 - After any number of changes, A can only go to either B or C but not both
 - After 100 changes, A can end up at B but not C
 - After 500 changes, A can end up at C but not B

What is a transition state diagram?

- A **transition diagram** is a **directed graph**
 - The **vertices** are the **states**
 - The **edges** represent the **transition probabilities** between the states
- The graph can contain
 - **Loops**
 - These will be the transition probabilities of the next state being the same as the current state



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- **Two edges between each pair** of vertices
 - The edges will be in opposite directions
 - Each edge will show the transition probability of the state changing in the given direction
- The **probabilities on the edges coming out** of a vertex **add up to 1**

 **Examiner Tip**

- Drawing a transition state diagram (even when the question does not ask for one) can help you visualise the problem

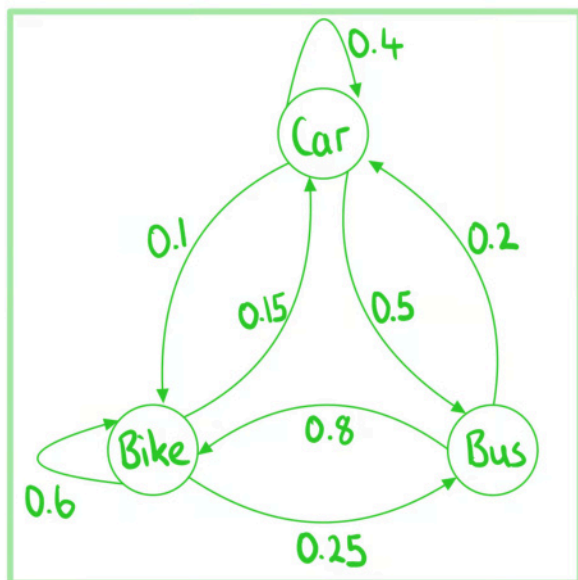
 **Worked example**

Fleur travels to work by car, bike or bus. Each day she chooses her mode of transport based on the transport she chose the previous day.

- If Fleur travels by car then there is a 40% chance that she will travel by car the following day and a 10% chance that she will travel by bike.
- If Fleur travels by bike then there is a 60% chance that she will travel by bike the following day and a 25% chance that she will travel by bus.
- If Fleur travels by bus then there is an 80% chance that she will travel by bike the following day and a 20% chance that she will travel by car.

Represent this information as a transition state diagram.

The probabilities on the arrows coming out of a state add to 1





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4.13.2 Transition Matrices

Transition Matrices

What is a transition matrix?

- A **transition matrix** T shows the **transition probabilities** between the current state and the next state
 - The **columns** represent the **current states**
 - The **rows** represent the **next states**
- The element of T in the i^{th} row and j^{th} column gives the transition probability t_{ij} of :
 - the **next state** being the state corresponding to **row** i
 - **given that the current state** is the state corresponding to **column** j
- The probabilities in each **column** must **add up to 1**
- The transition matrix depends on how you assign the states to the columns
 - Each transition matrix for a Markov chain will contain the same elements
 - The rows and columns may be in different orders though
 - E.g. Sunny (S) & Cloudy (C) could be in the order **S then C** or **C then S**

What is an initial state probability matrix?

- An **initial state probability matrix** s_0 is a column vector which contains the **probabilities** of each state being chosen as the **initial state**
 - If you know which state was chosen as the initial state then that entry will be 1 and the others will all be zero
- You can find the **state probability matrix** s_1 which contains the probabilities of each state being chosen after **one interval of time**
 - $s_1 = Ts_0$

How do I find expected values after one interval of time?

- Suppose the Markov change represents a **population moving between states**
 - Examples include:
 - People in a town switching gyms each year
 - Children choosing a type of sandwich for their lunch each day
- Suppose the **total population is fixed** and equals N
- You can **multiply the state probability matrix** s_1 by N to find the expected number of members of the population at each state

Examiner Tip

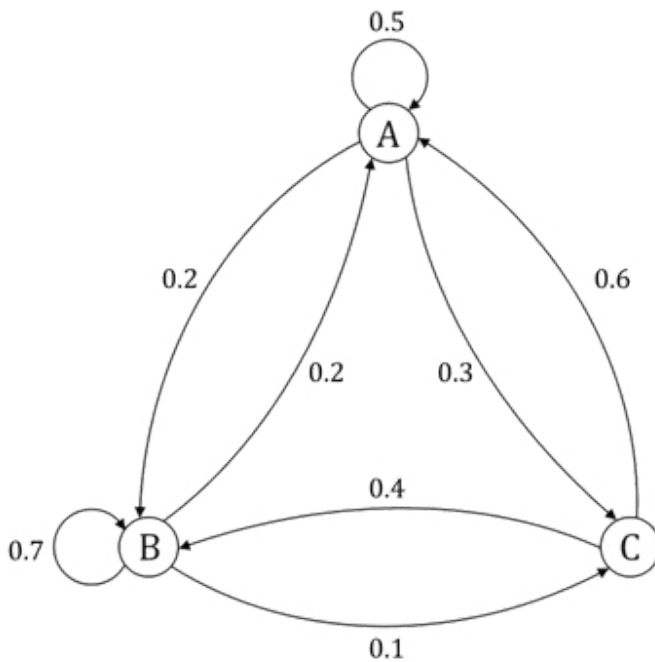
- If you are asked to find a transition matrix, check that all the probabilities within a column add up to 1
- Drawing a transition state diagram can help you to visualise the problem



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 **Worked example**

Each year Jamie donates to one of three charities: A, B or C. At the start of each year, the probabilities of Jamie continuing donate to the same charity or changing charities are represented by the following transition state diagram:



a) Write down a transition matrix T for this system of probabilities.

Current state

	A	B	C
Next state	A	B	C

$$T = \begin{pmatrix} 0.5 & 0.2 & 0.6 \\ 0.2 & 0.7 & 0.4 \\ 0.3 & 0.1 & 0 \end{pmatrix}$$

b) There is a 10% chance that charity A is the first charity that Jamie chooses, a 10% chance for charity B and an 80% chance for charity C. Find the charity which has the highest probability of being picked as the second charity after the first year.



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Write down the initial state vector $s_0 = \begin{pmatrix} 0.1 \\ 0.1 \\ 0.8 \end{pmatrix}$

$$s_1 = T s_0 \quad s_1 = \begin{pmatrix} 0.5 & 0.2 & 0.6 \\ 0.2 & 0.7 & 0.4 \\ 0.3 & 0.1 & 0 \end{pmatrix} \begin{pmatrix} 0.1 \\ 0.1 \\ 0.8 \end{pmatrix} = \begin{pmatrix} 0.55 \\ 0.41 \\ 0.04 \end{pmatrix}$$

Charity A has the highest probability of being the second charity picked.



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Powers of Transition Matrices

How do I find powers of a transition matrix?

- You can simply use your **GDC** to find given powers of a matrix
- The power could be left in terms of an **unknown** n
 - In this case it would be more helpful to write the transition matrix in diagonalised form (see section **1.8.2 Applications of Matrices**) $T = PDP^{-1}$ where
 - D is a **diagonal matrix** of the **eigenvalues**
 - P is a matrix of **corresponding eigenvectors**
 - Then $T^n = PD^nP^{-1}$
 - This is given in the **formula booklet**
 - Every transition matrix always has an **eigenvalue equal to 1**

What is represented by the powers of a transition matrix?

- The powers of a transition matrix also **represent probabilities**
- The element of T^n in the i^{th} row and j^{th} column gives the **probability** t_{ij}^n of:
 - the **future state** after **n intervals of time** being the state corresponding to **row i**
 - **given that** the **current state** is the state corresponding to **column j**
- For example: Let T be a transition matrix with the element $t_{2,3}$ representing the probability that tomorrow is sunny given that it is raining today
 - The element $t_{2,3}^5$ of the matrix T^5 represents the probability that it is sunny in 5 days' time given that it is raining today
- The probabilities in **each column** must still **add up to 1**

How do I find the column state matrices?

- The column state matrix s_n is a column vector which contains the **probabilities** of each state being chosen after n intervals of time given the current state
 - s_n depends on s_0
- To calculate the column state matrix you raise the transition matrix to the power n and multiply by the initial state matrix
 - $T^n s_0 = s_n$
 - You are given this in the **formula booklet**
- You can multiply s_n by the fixed population size to find the expected number of members of the population at each state after n intervals of time



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Worked example

At a cat sanctuary there are 1000 cats. If a cat is brushed on a given day, then the probability it is brushed the following day is 0.2. If a cat is not brushed on a given day, then the probability that it will be brushed the following day is 0.9.

The transition matrix T is used to model this information with $T = \begin{pmatrix} 0.2 & 0.9 \\ 0.8 & 0.1 \end{pmatrix}$.

- a) On Monday Hippo the cat is brushed. Find the probability that Hippo will be brushed on Friday.

Identify the states with the rows/columns

$$\begin{array}{c} \text{Next} \\ \begin{matrix} B \\ B' \end{matrix} \end{array} \begin{array}{c} \text{Current} \\ \begin{matrix} B & B' \end{matrix} \end{array} \begin{pmatrix} 0.2 & 0.9 \\ 0.8 & 0.1 \end{pmatrix}$$

Friday is 4 days after Monday

$$T^4 = \begin{pmatrix} 0.2 & 0.9 \\ 0.8 & 0.1 \end{pmatrix}^4 = \begin{pmatrix} 0.6424 & 0.4023 \\ 0.3576 & 0.5977 \end{pmatrix} \begin{array}{l} B \\ B' \end{array} \left. \vphantom{\begin{pmatrix} 0.6424 & 0.4023 \\ 0.3576 & 0.5977 \end{pmatrix}} \right\} \text{Future}$$

$\underbrace{\hspace{10em}}_{\text{Current}}$

Current = B
 Future = B

0.6424

- b) On Monday 700 cats were brushed. Find the expected number of cats that will be brushed on the following Monday.



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On Monday 700 brushed $s_0 = \begin{pmatrix} 0.7 \\ 0.3 \end{pmatrix}$

Expected numbers after 7 days

$$\text{Total} \times S_7 = \text{Total} \times T^7 s_0$$

$$1000 \times \begin{pmatrix} 0.2 & 0.9 \\ 0.8 & 0.1 \end{pmatrix}^7 \begin{pmatrix} 0.7 \\ 0.3 \end{pmatrix} = \begin{pmatrix} 0.2 & 0.9 \\ 0.8 & 0.1 \end{pmatrix}^7 \begin{pmatrix} 700 \\ 300 \end{pmatrix} = \begin{pmatrix} 515.36309 \\ 484.63691 \end{pmatrix} \begin{matrix} B \\ B' \end{matrix}$$

515 cats



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Steady State & Long-term Probabilities

What is the steady state of a regular Markov chain?

- The vector \mathbf{s} is said to be a **steady state** vector if it does not change when multiplied by the transition matrix
 - $T\mathbf{s} = \mathbf{s}$
- **Regular Markov chains** have steady states
 - A Markov chain is said to be regular if there exists a **positive integer k** such that **none of the entries are equal to 0** in the matrix T^k
 - For this course all Markov chains will be regular
- The transition matrix for a regular Markov chain will have **exactly one** eigenvalue equal to 1 and the **rest will all be less than 1**
- As n gets bigger T^n tends to a matrix where **each column is identical**
 - The column matrix formed by using **one of these columns** is called the steady state column matrix \mathbf{s}
 - This means that the **long-term probabilities** tend to fixed probabilities
 - s_n tends to \mathbf{s}

How do I use long-term probabilities to find the steady state?

- As T^n tends to a matrix whose columns equal the steady state vector
 - Calculate T^n for a large value of n using your GDC
 - If the columns are identical when rounded to a required degree of accuracy then the column is the steady state vector
 - If the columns are not identical then choose a higher power and repeat

How do I find the exact steady state probabilities?

- As $T\mathbf{s} = \mathbf{s}$ the steady state vector \mathbf{s} is the **eigenvector** of T corresponding to the **eigenvalue equal to 1** whose elements sum to 1:
 - Let \mathbf{s} have entries x_1, x_2, \dots, x_n
 - Use $T\mathbf{s} = \mathbf{s}$ to form a system of linear equations
 - There will be an infinite number of solutions so choose a value for one of the unknowns
 - For example: let $x_n = 1$
 - Ignoring the last equation solve the system of linear equations to find x_1, x_2, \dots, x_{n-1}
 - Divide each value x_i by the sum of the values
 - This makes the values add up to 1
- You might be asked to **show this result using diagonalisation**
 - Write $T = PDP^{-1}$ where D is the diagonal matrix of eigenvalues and P is the matrix of eigenvectors
 - Use $T^n = PD^nP^{-1}$
 - As n gets large D^n tends to a matrix where all entries are 0 apart from one entry of 1 due to the eigenvalue of 1
 - Calculate the limit of T^n which will have **identical columns**
 - You can calculate this by multiplying the three matrices (P, D^∞, P^{-1}) together

Examiner Tip

- If you calculate T^∞ by hand then a quick check is to see if the columns are identical

- It should look like
$$\begin{pmatrix} a & a & a \\ b & b & b \\ c & c & c \end{pmatrix}$$



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Worked example

If a cat is brushed on a given day, then the probability it is brushed the following day is 0.2. If a cat is not brushed on a given day, then the probability that it will be brushed the following day is 0.9.

The transition matrix T is used to model this information with $T = \begin{pmatrix} 0.2 & 0.9 \\ 0.8 & 0.1 \end{pmatrix}$.

- a) Find an eigenvector of T corresponding to the eigenvalue 1.

\underline{v} is an eigenvector of T with eigenvalue 1 if $T\underline{v} = \underline{v}$

Let $\underline{v} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$

$$T\underline{v} = \begin{pmatrix} 0.2 & 0.9 \\ 0.8 & 0.1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0.2x_1 + 0.9x_2 \\ 0.8x_1 + 0.1x_2 \end{pmatrix}$$

$$T\underline{v} = \underline{v} \quad 0.2x_1 + 0.9x_2 = x_1 \Rightarrow 0.9x_2 = 0.8x_1 \Rightarrow 9x_2 = 8x_1$$

$$0.8x_1 + 0.1x_2 = x_2 \Rightarrow 0.8x_1 = 0.9x_2 \Rightarrow 8x_1 = 9x_2$$

Find a solution $x_1 = 9$ and $x_2 = 8$

$\begin{pmatrix} 9 \\ 8 \end{pmatrix}$ or any scalar multiple

- b) Hence find the steady state vector.

Scale the elements so that they add to 1 $\begin{pmatrix} 9 \\ 17 \\ 8 \\ 17 \end{pmatrix}$

The eigenvector corresponding to the eigenvalue 1, whose elements add to 1, is the steady state vector.

$$\begin{pmatrix} 9 \\ 17 \\ 8 \\ 17 \end{pmatrix}$$



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