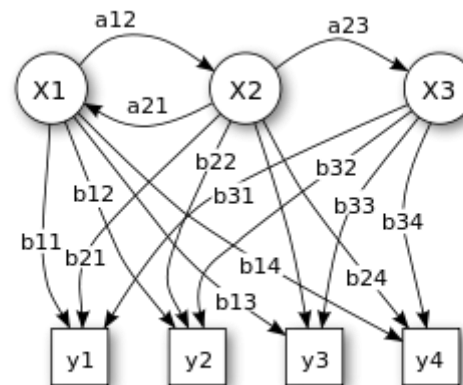


Markov Models

Applied to Anomaly Detection

By Saab Group





The paper



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Anomaly detection based on a dynamic Markov model

Huorong Ren ^{a, b}, Zhixing Ye ^{a, b}✉, Zhiwu Li ^{c, a}

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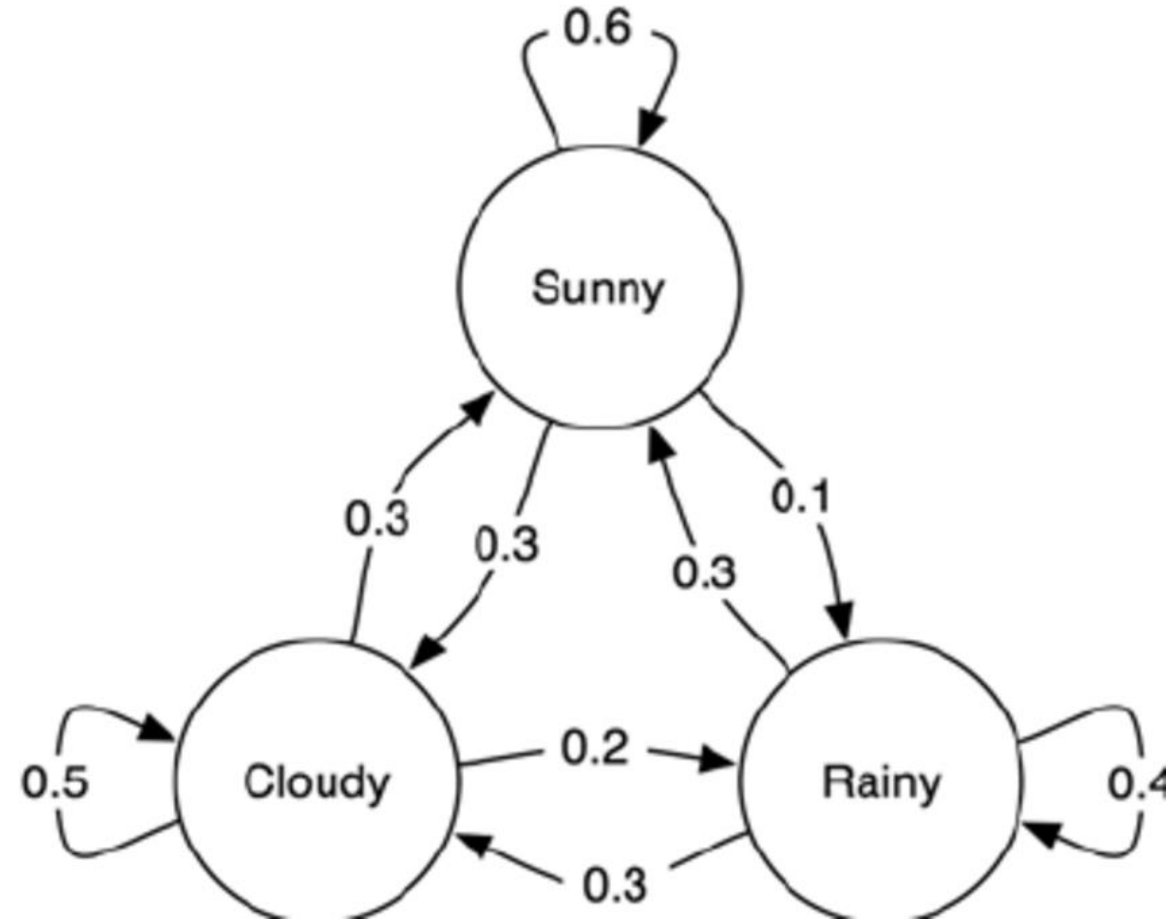
What is a Markov Model?

- In probability theory, a Markov model is a stochastic model **used to model randomly changing systems.**
- It is assumed that **future states depend only on the current state**, not on the events that occurred before it (that is, it assumes the Markov property).

What is a Markov Model?

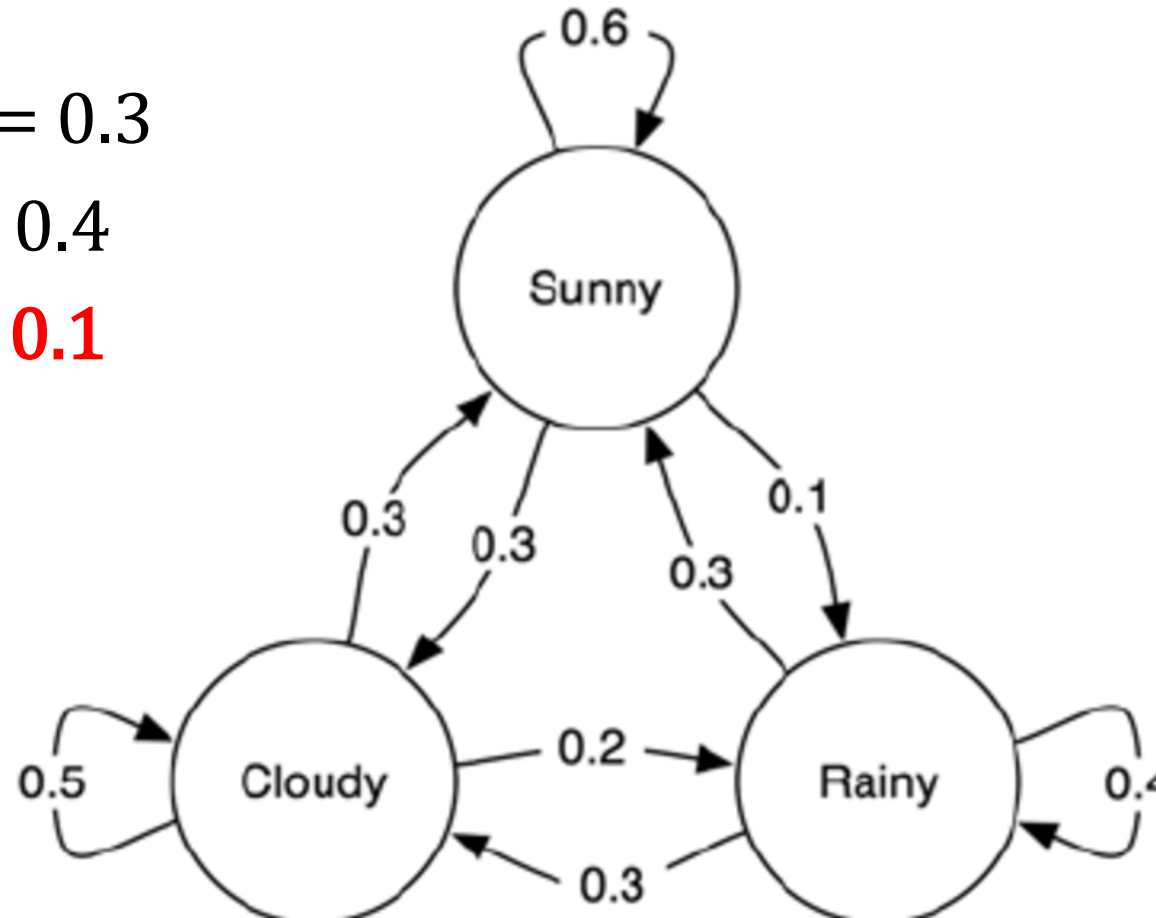


What is a Markov Model?



Why is this helpful in anomaly detection?

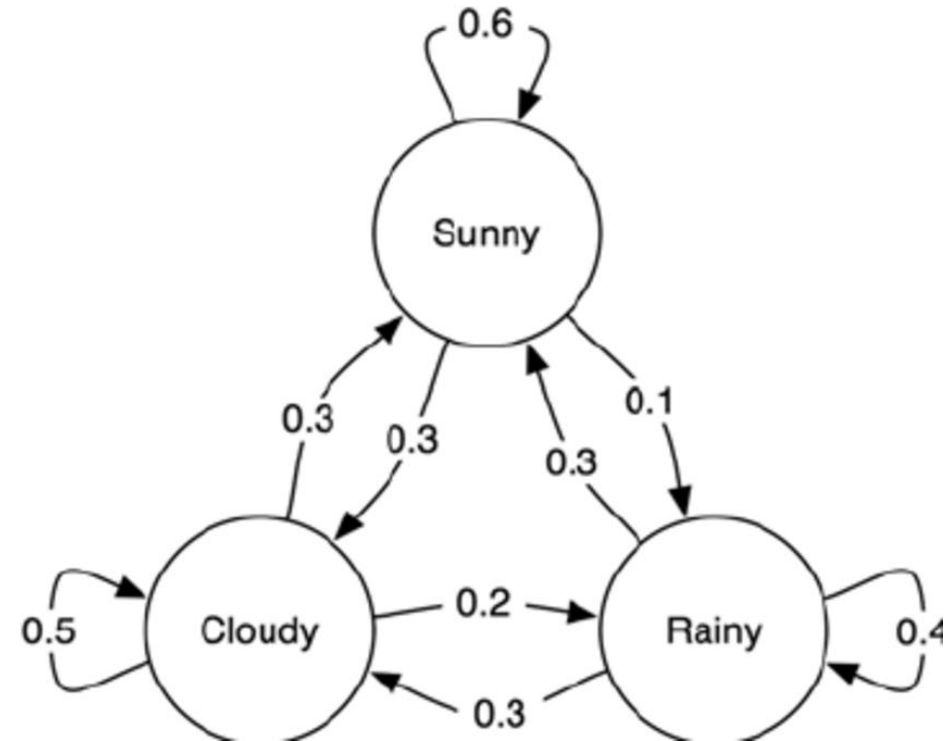
- $P(\text{sunny}|\text{cloudy}) = 0.3$
- $P(\text{rainy} | \text{rainy}) = 0.4$
- $P(\text{rainy}|\text{sunny}) = \mathbf{0.1}$



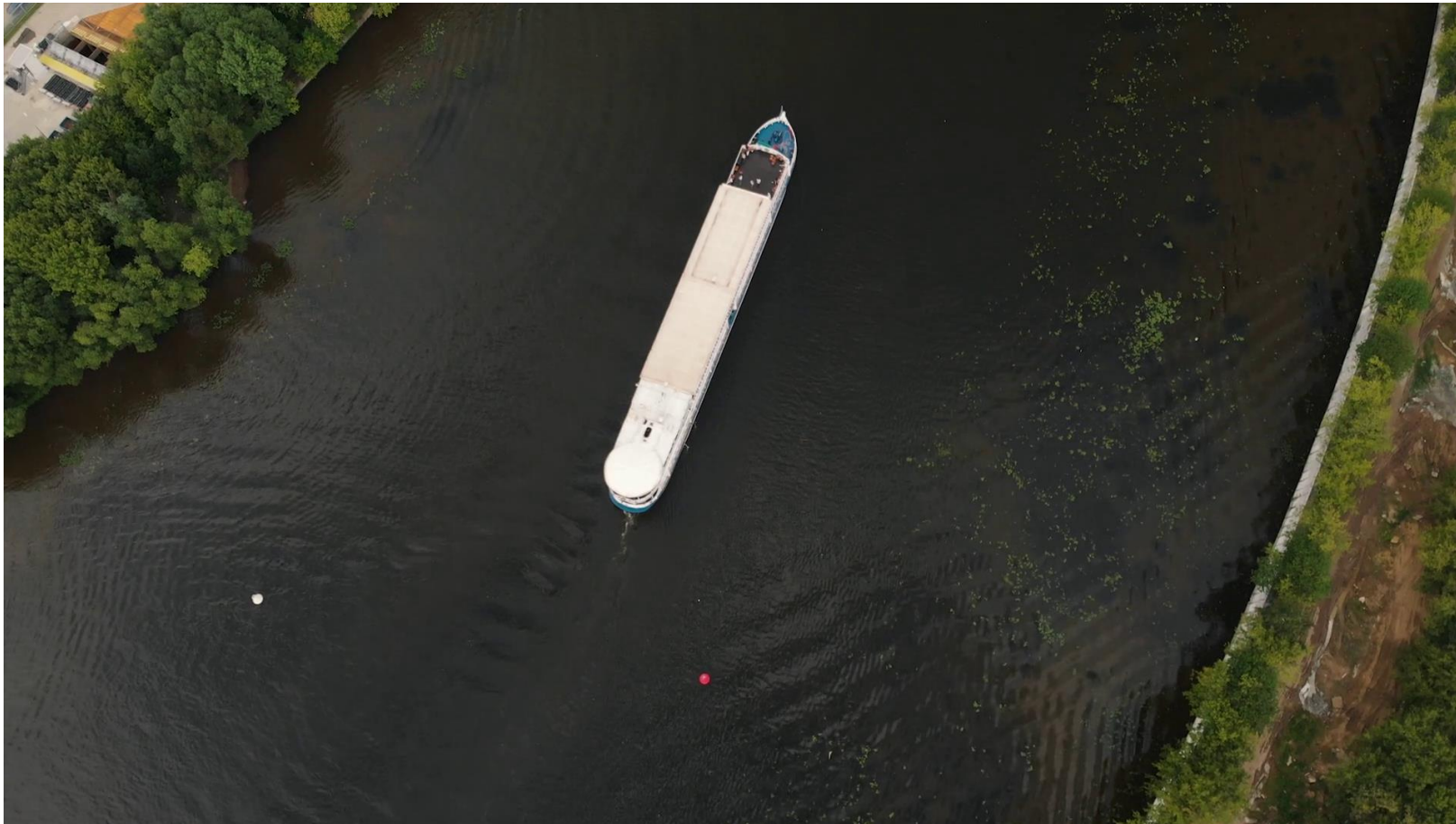
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↓
ANOMALY
DETECTED



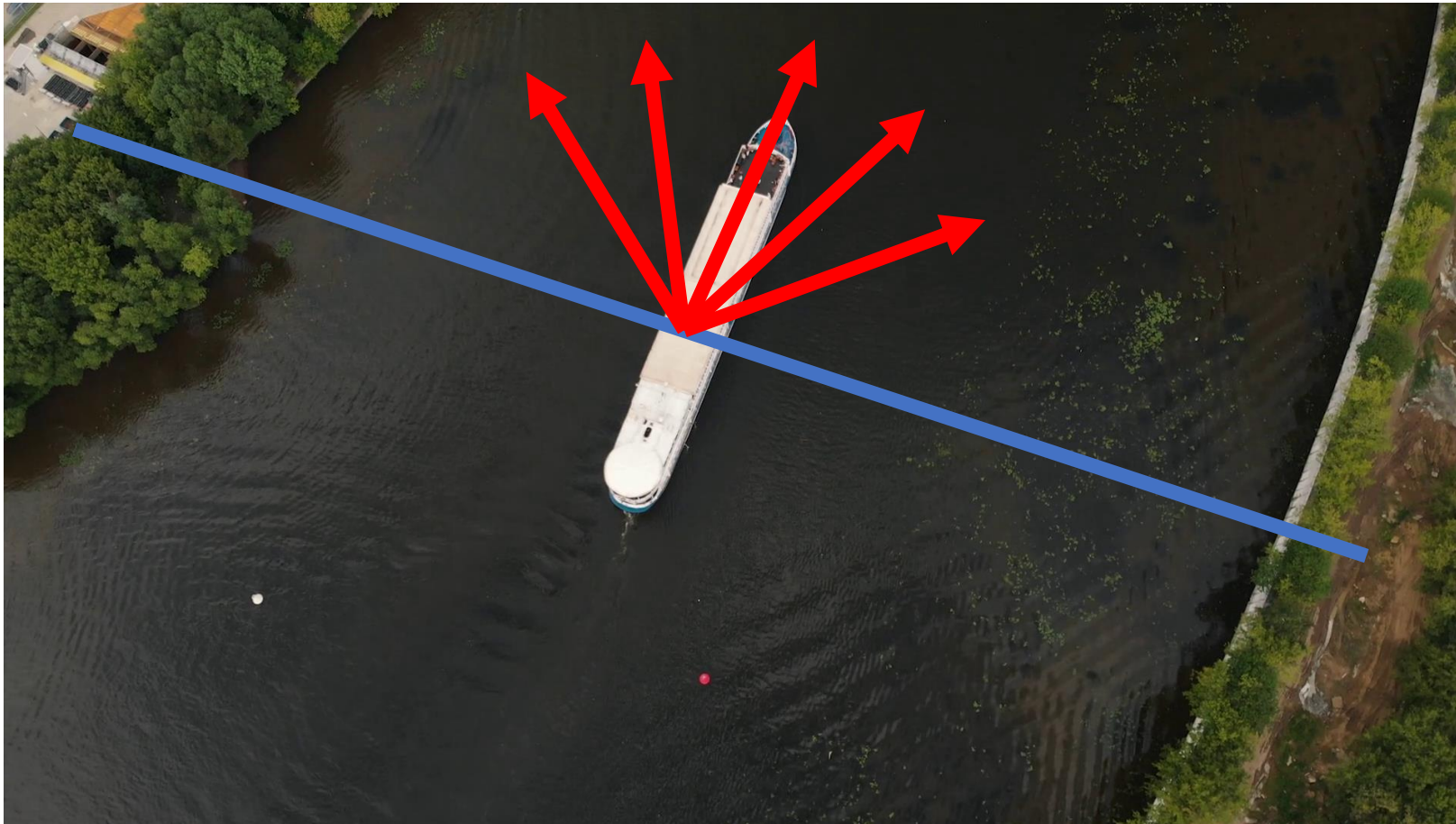
Why can this be applied to our Project?



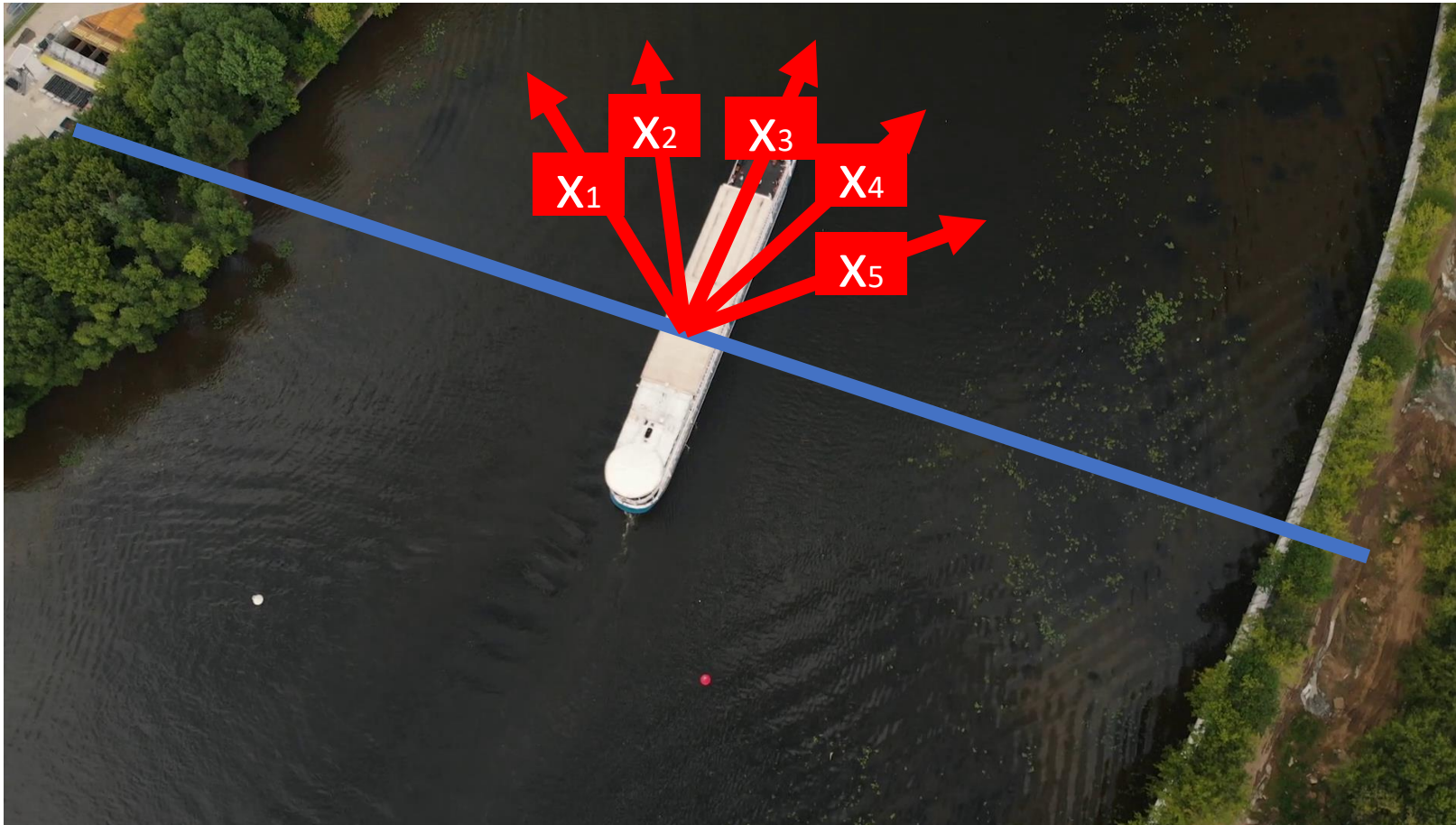
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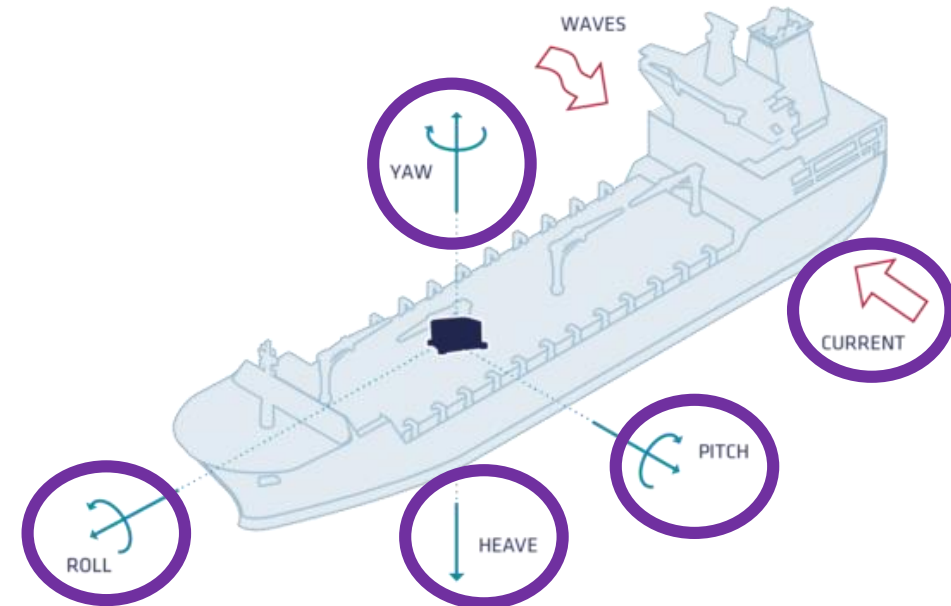
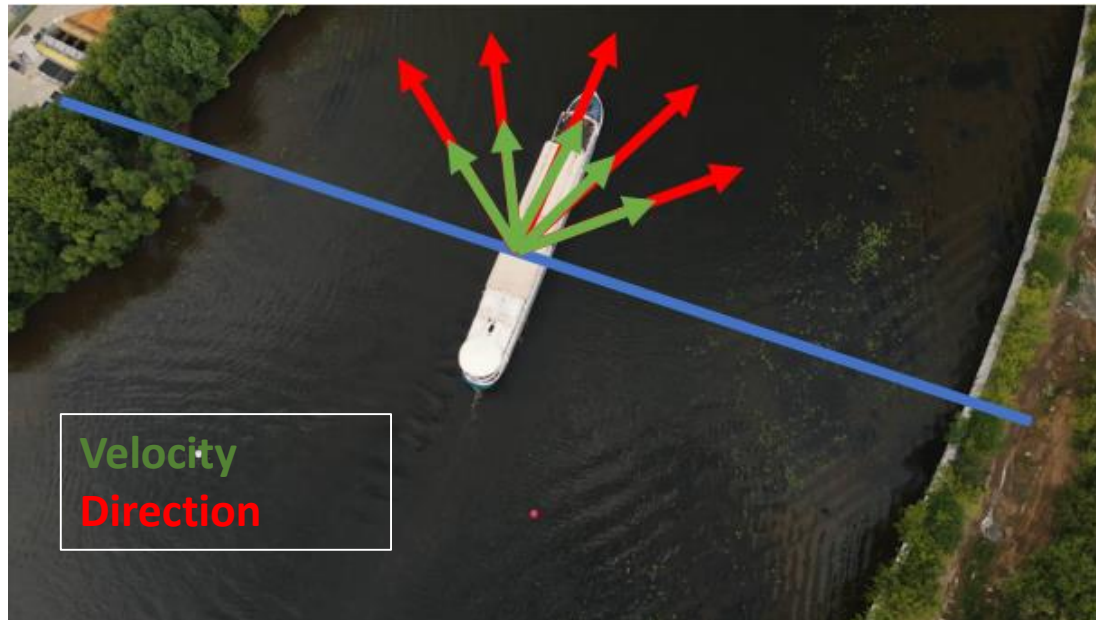
Why can this be applied to our Project?





Transitions

And... if we have other interesting features?



Since all measurements are bounded

- We can partitionate them and classify them
- E.g:

So what's the average speed of a sailboat? **Most sailboats cruise at a speed of 4-6 knots (4.5-7 mph), with a top speed of 7 knots (8 mph or 13 km/h).** Larger racing yachts can easily reach speeds up to 15 knots (17 mph or 28 km/h), with an average cruising speed between 6-8 knots (7-9 mph). Cruising speeds of over 8 knots are uncommon.

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0 km/h

30 km/h

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Since all measurements are bounded

- We can partitionate them and classify them.
- Now each classification is a substate of the measurement, so that for speed we have both **slow**, **normal** and **fast** sub-states.
- Position could be partitioned into **suspicious** and **non-suspicious** according to the history of anomalies detected in a certain port.
- The direction can also be partitioned (we Will see it in the following slides).



Since all measurements are bounded

- So if we use in our model 3 states representing the speed, 2 states representing the position and 5 states representing the direction, there will be:

$$\binom{10}{3} = 120 \text{ states}$$



Since all measurements are bounded

- So if we use in our model 3 states representing the speed, 2 states representing the position and 5 states representing the direction, there will be:

$$\binom{10}{3} = 120 \text{ states}$$

- This might be too much (or not)





More Data

=

**More information
per state**

An example focusing only in direction

- We can extract much more features (depends on the data)

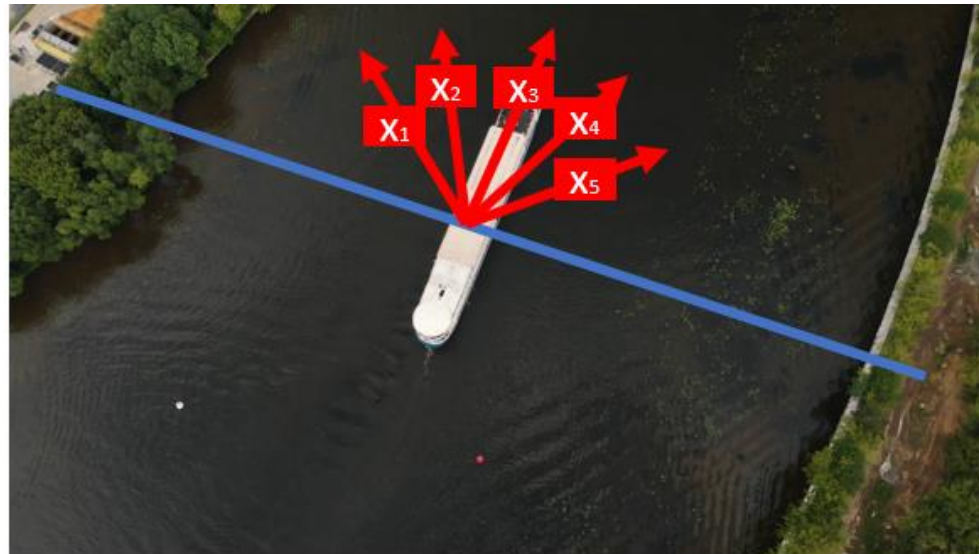
x1

x2

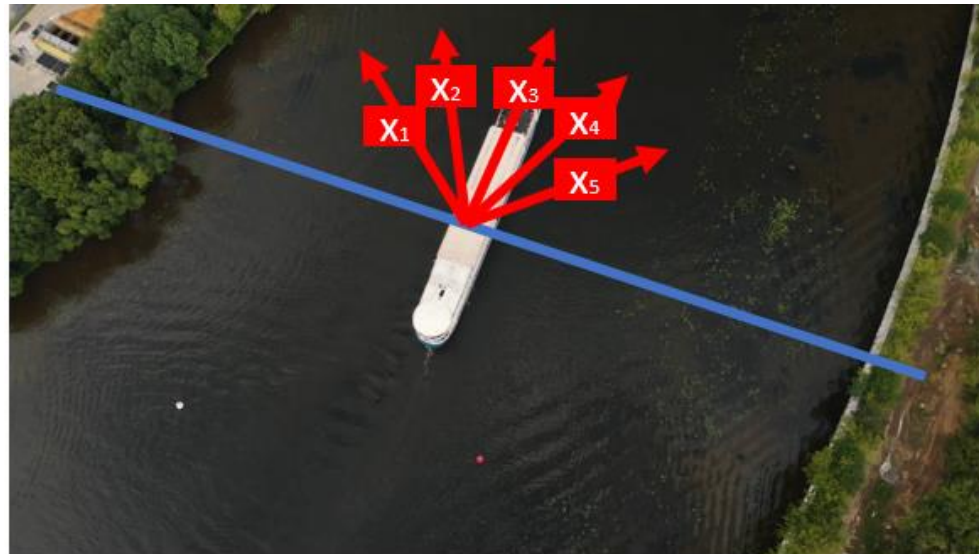
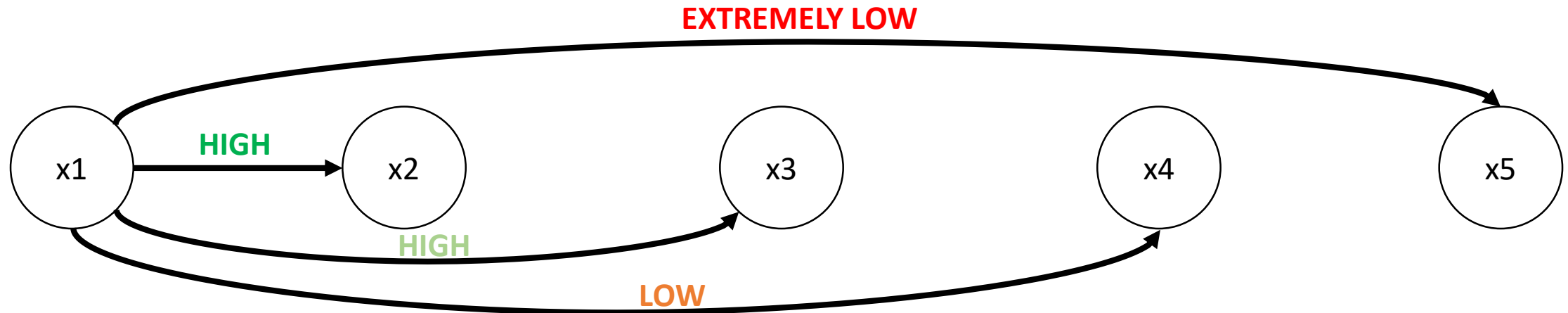
x3

x4

x5

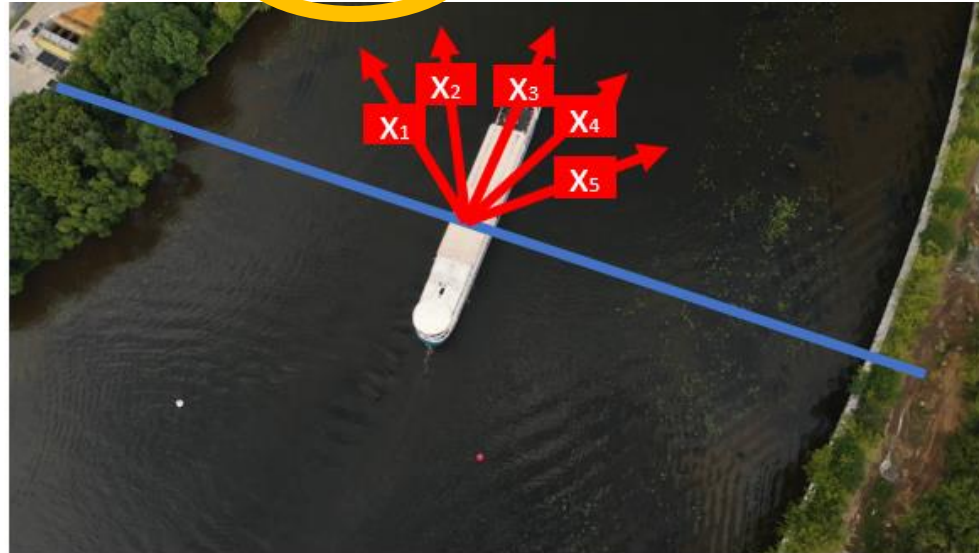
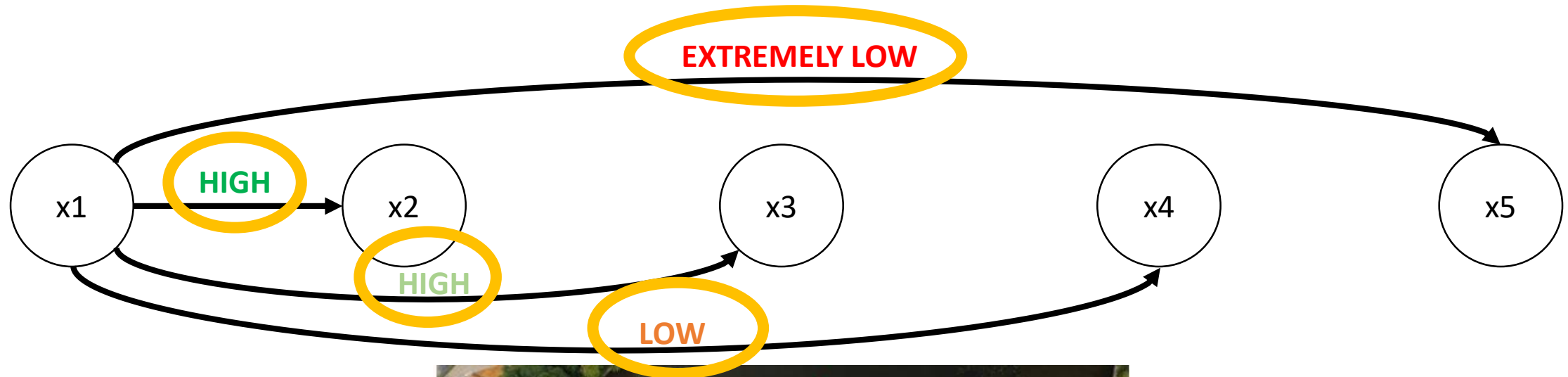


Basic Model



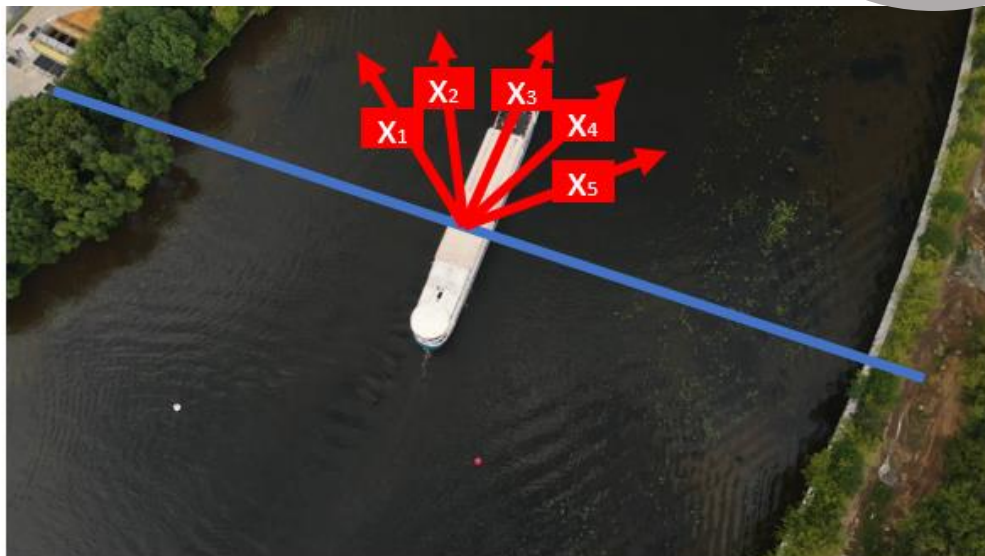
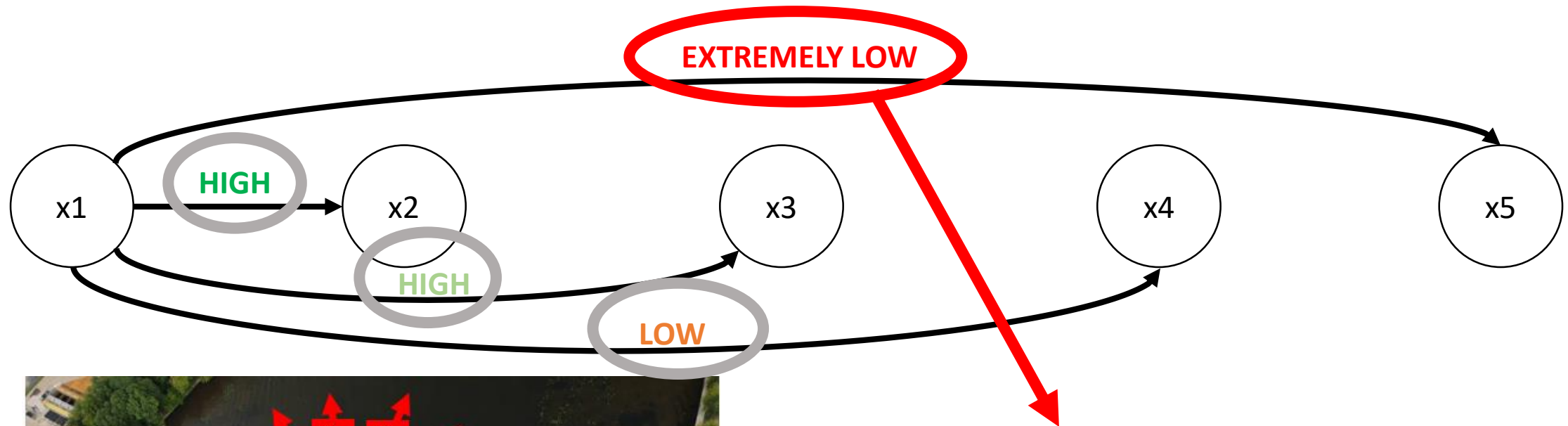
Basic Model

What could we extract from the _____



Basic Model

Finally...



If there is a transition from x1 to x5:

WE ARE JUST IN FRONT OF AN

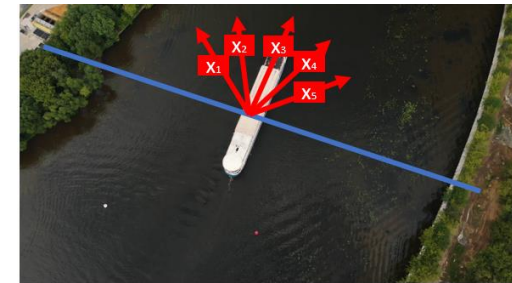
Implementation

Finally...

Once defined the states, we can compute probabilities of transition between one state to another and represent them on a matrix

Transition Probability Matrix

$$\begin{matrix}
 & \begin{matrix} 1 & 2 & \dots & n \end{matrix} \\
 \begin{matrix} 1 \\ 2 \\ 3 \\ \vdots \\ m \end{matrix} & \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ a_{31} & a_{32} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}
 \end{matrix}$$



a_{ij} = Probability of transition from i to j

Idea:

- Get the probability values from a training set
- Evaluate the model with the test set
 - Whenever a transition happen, check the probability of that to have occurred and if it is under a tested threshold → **ANOMALY**

Also Interesting

❑ Problem

❑ Speed and size of the ships should get evaluated in conjunction.

- Higher speeds are supposed to be adopted by smaller ships
- High speeds are not anomalous for big ships

❑ Markov Models allow us to evaluate speed and length together

❑ Given a range of possible values for speed and length, we can partitionate both so that given a transition pair

$(\text{speed1}, \text{length}) \rightarrow (\text{speed1}, \text{length})$

Initially: $(100\text{km/h}, 500\text{m}) \rightarrow \rightarrow (\text{type } 7, \text{type } 5) \rightarrow \text{STATE } 4$

Finally: $(20 \text{ km/h}, 500\text{m}) \rightarrow \rightarrow (\text{type } 4, \text{type } 5) \rightarrow \text{STATE } 7$

SO

STATE 4 to STATE 7

