

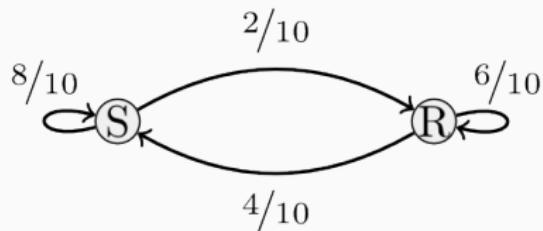
# ACTL2102 Week 7 - CTMC: Kolmogorov Equations

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Tadhg Xu-Glassop

2025T3

## Revision of DTMC's



Recall the Discrete Time Markov Chain model - at each step, we have probabilities of transitioning to other states, or remaining in our current state.

## Revision of Poisson Process

Further recall our definition of the Poisson process,

### Definition (Poisson Process)

A counting process  $\{N(t) : t \geq 0\}$  is said to be a Poisson process if:

1.  $N(0) = 0$ ;
2. it has independent increments;
3. it has stationary increments;
4. for  $h \rightarrow 0$ ;

$$\mathbb{P}(N(t+h) - N(t) = 1) = \lambda h + o(h),$$

and

$$\mathbb{P}(N(t+h) - N(t) > 1) = o(h)$$

## Motivating CTMC's

We want to extend our discrete model where we can define our own state space and transition possibilities to a continuous case; transition can occur at any time, not just periodically.

We can do this by generalising our Poisson process.

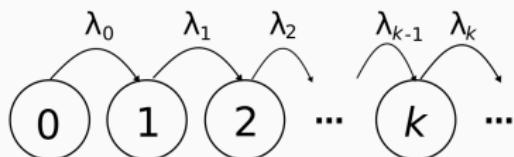
## Poisson Process is a CTMC

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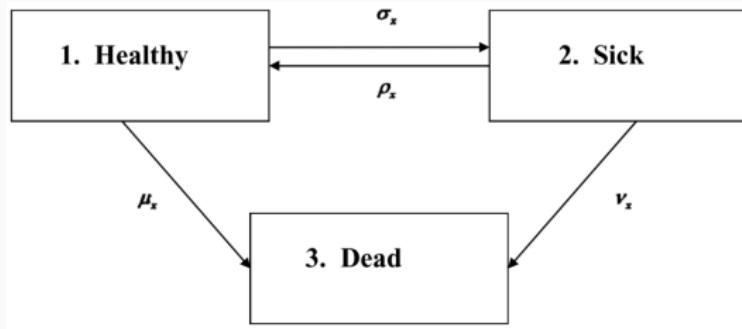
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We can visualise it in a very similar way to DTMC's. Recall in a very small amount of time  $h$ , the probability we transition to the next state is  $\lambda$ . So, we can represent these infinitesimal transition probabilities with a very similar diagram.



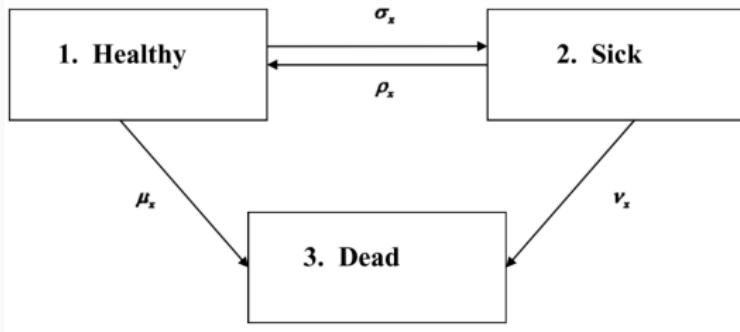
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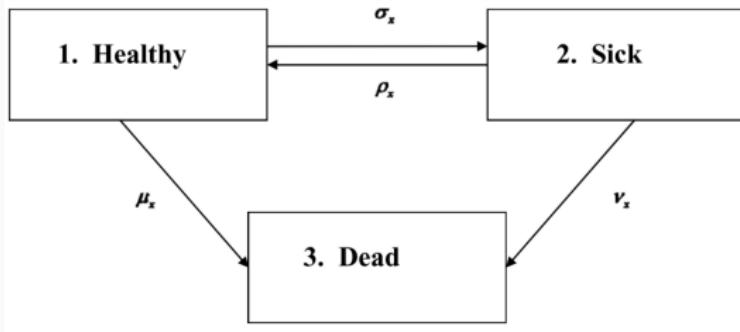
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Consider the Healthy state - in a small amount of time  $h$ , we have a  $\sigma$  and  $\mu$  probability of transitioning to sick and dead respectively.

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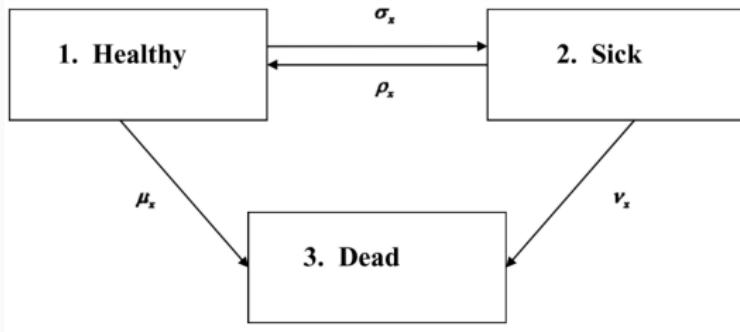


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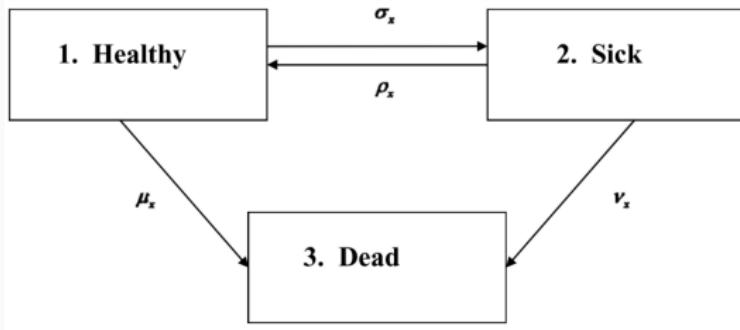


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Consider the Healthy state - in a small amount of time  $h$ , we have a  $\sigma$  and  $\mu$  probability of transitioning to sick and dead respectively.

- What is the total probability of transitioning out of healthy in a small interval  $h$ ? A:  $\sigma + \mu$ .
- This is the definition of exponential distribution  $\implies$  the time we spend in Healthy is  $\text{Exp}(\mu + \sigma)$ .

# Transition Probabilities

## Definition

For a continuous time Markov chain  $X$ , define

$$P_{ij}(s, s + t) = \mathbb{P}(X(s + t) = j \mid X(s) = i).$$

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Let  $q_{ij}$  be the instantaneous probability we transition from state  $i$  to  $j$  (as before). Then,

$$\frac{\partial}{\partial t} P_{ij}(s, s + t) = \frac{\partial}{\partial t} P_{ij}(t) = q_{ij}.$$

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Intuitively, recall the rate we leave state  $i$  is the sum of all rates. So, the rate we stay in the state is the negative of this.

## Motivinating CK Equations

So its clear that  $q_{ij}$  are very important and essentially define a CTMC and by definition,  $\frac{\partial}{\partial t} P_{ij}(t) = q_{ij}$  so we just need to compute this!

But how do we find this...

# Kolmogorov Equations

## Theorem (Kolmogorov Equations)

For a homogenous Markov chain with state space  $S$ , for  $t \geq 0$  we have

- Kolmogorov's Forward Equation

$$\frac{\partial}{\partial t} P_{ij}(t) = \sum_{k \in S} P_{ik}(t) q_{kj},$$

- Kolmogorov's Backward Equation

$$\frac{\partial}{\partial t} P_{ij}(t) = \sum_{k \in S} q_{ik} P_{kj}(t).$$

So, we can solve for our transition probabilities by solving these differential equations, subject to the initial condition  $P_{ij}(0) = 0$  if  $i \neq j$  and 1 if  $i = j$ .