## Multi-state Modeling

The introduction and practice

Amarit Tansawet, MD, PhD. 15/11/2024 Sukho Place Building





## Multi-state working group











... and more to come

## Recap

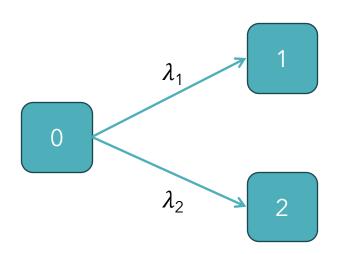
Hazard function

$$h(t) = \lim_{\Delta t \downarrow 0} \frac{P(t \le T < t + \Delta t \mid T \ge t)}{\Delta t}$$

### Recap

Cause-specific hazard

$$\lambda_k(t) = \lim_{\Delta t \downarrow 0} \frac{P(t \le T < t + \Delta t, E = k \mid T \ge t)}{\Delta t}$$



$$\sum_{e=1}^{K} \lambda_e(t) = \lim_{\Delta t \downarrow 0} \frac{\sum_{e=1}^{K} P(t \le T < t + \Delta t, E = e \mid T \ge t)}{\Delta t}$$
$$= \lim_{\Delta t \downarrow 0} \frac{P(t \le T < t + \Delta t \mid T \ge t)}{\Delta t} = h(t).$$

#### Multi-state model

Initial state

#0001

Bulbasaur

Grass · Poison

Transit 1

Intermediate state

Absorbing state







**Ivysaur** 

Grass · Poison



Remain in the same state

Venusaur

Grass · Poison

Mutually exclusive

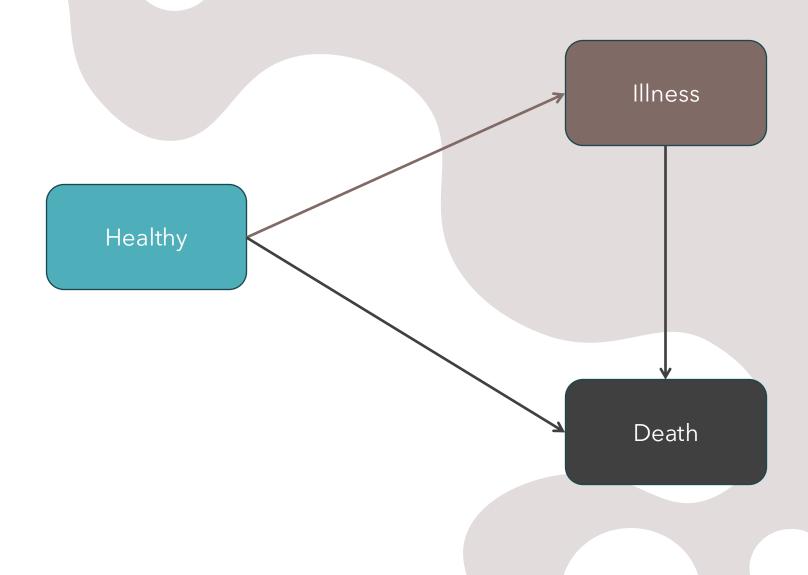
## Why do we need multi-state model?

Need information on etiological mechanism

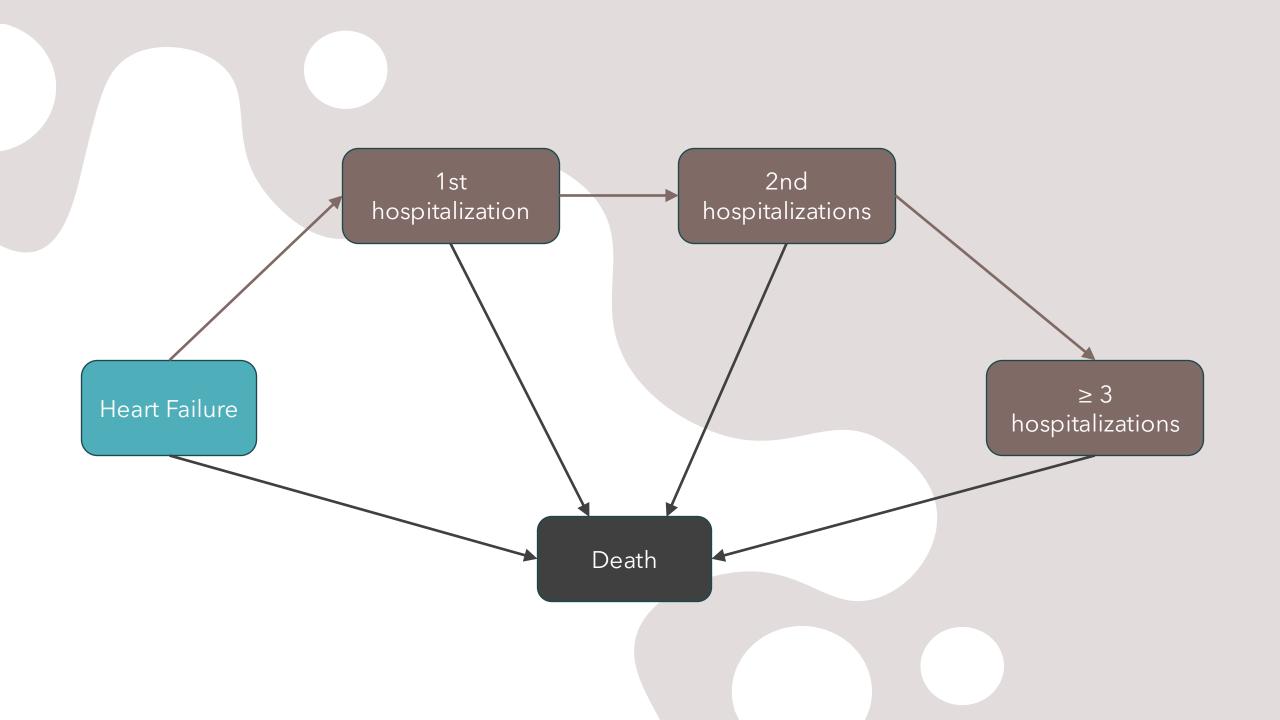
More accurate prediction

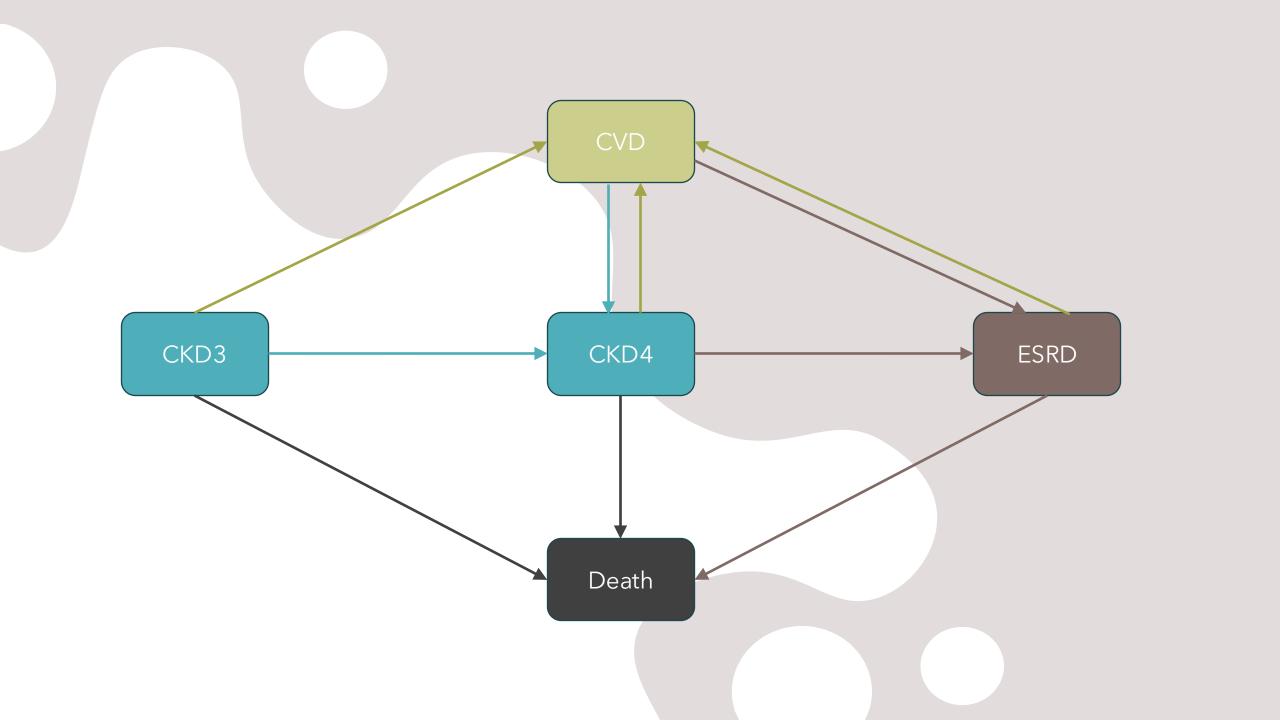
Patients don't have to enter from the initial state

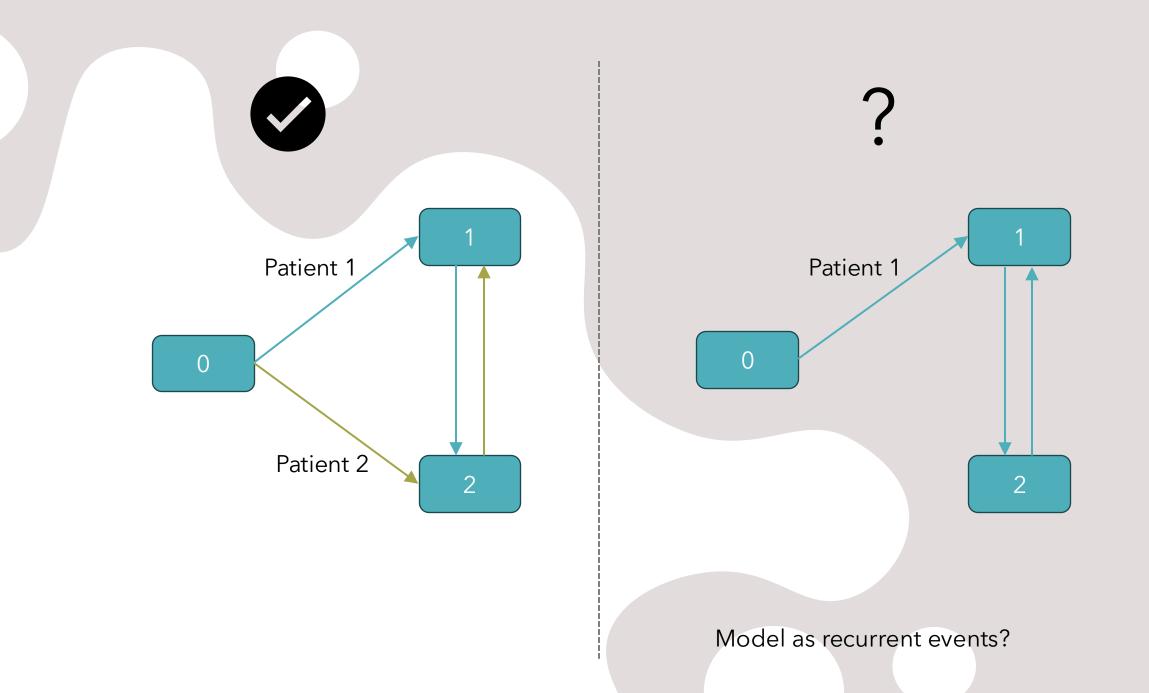
#### Illness-death model



ACYCLIC





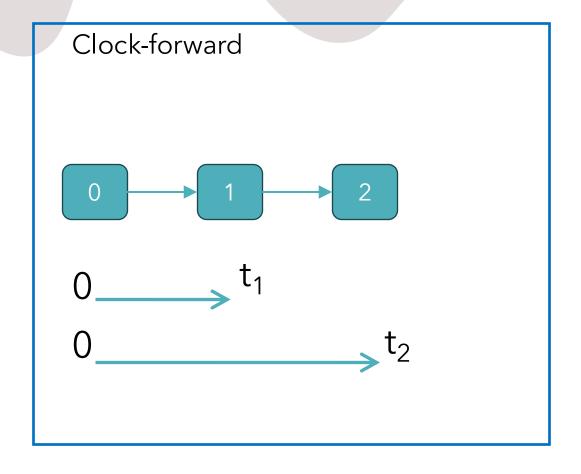


## Markov property

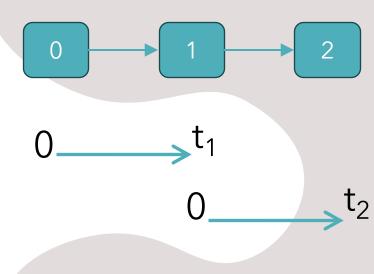
What happens next only depends on the current state, not on what happened before

Need to be relaxed in more complex models

#### Time scale



Clock-reset



Model different transitions with different time scales?

## Basic data format: Stage-ordered format

ID	1.time	1.status	2.time	2.status	3.time	3.status
1	10	1	20	0	20	0
2	20	0	20	0	20	0
3	10	1	20	1	30	0
4	10	1	20	1	30	1
5	10	1	20	0	20	1
6	20	0	20	0	20	1

#### Basic data format

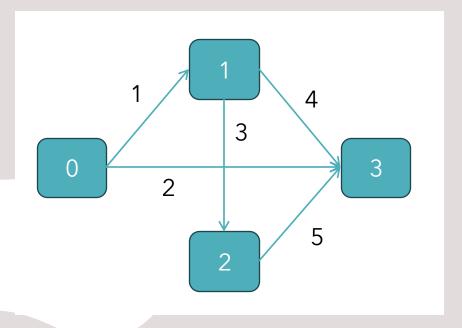
Stage-ordered format (irreversible multi-state data): One column contains the first observation time in that state and the other column is a zero-one variable that denotes whether the state was reached during follow-up

## Basic data format: Stage-ordered format

1 10	4			
	1 20	0	20	0
2 20	0 20	0	20	0
3 10 Sta	ate diac	ara¹m'	<b>7</b> 30	0
4 10	1 20	1	30	1
5 10	1 20	0	20	1
6 20	0 20	0	20	1

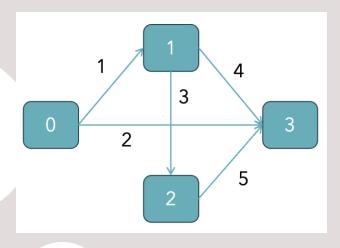
## Basic data format: Stage-ordered format

ID	1.time	1.status	2.time	2.status	3.time	3.status
1	10	1	20	0	20	0
2	20	0	20	0	20	0
3	10	1	20	1	30	0
4	10	1	20	1	30	1
5	10	1	20	0	20	1
6	20	0	20	0	20	1



ID	1.time	1.status	2.time	2.status	3.time	3.status
1	10	1	20	0	20	0

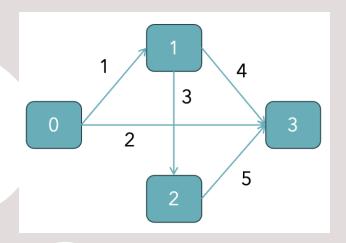
ID	From	То	Trans	Start	Stop	Status
1	0	1	1	0	10	1
1	0	3	2	0	10	0
1	1	2	3	10	20	0
1	1	3	4	10	20	0



i.e., 
$$1 \rightarrow censor$$

ID	1.time	1.status	2.time	2.status	3.time	3.status
2	20	0	20	0	20	0

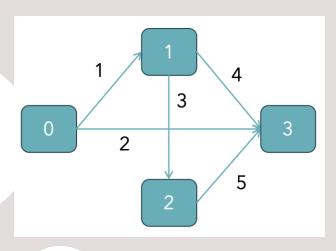
ID	From	То	Trans	Start	Stop	Status
2	0	1	1	0	20	0
2	0	3	2	0	20	0



i.e., 0→ censor

ID	1.time	1.status	2.time	2.status	3.time	3.status
3	10	1	20	1	30	0

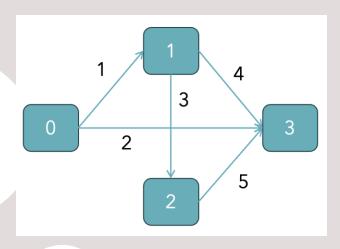
ID	From	То	Trans	Start	Stop	Status
3	0	1	1	0	10	1
3	0	3	2	0	10	0
3	1	2	3	10	20	1
3	1	3	4	10	20	0
3	2	3	5	20	30	0



i.e.,  $2 \rightarrow$  censor

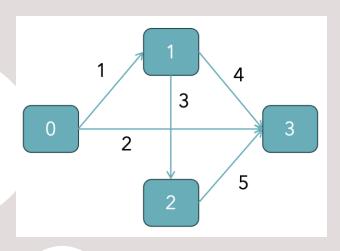
ID	1.time	1.status	2.time	2.status	3.time	3.status
4	10	1	20	1	30	1

ID	From	То	Trans	Start	Stop	Status
4	0	1	1	0	10	1
4	0	3	2	0	10	0
4	1	2	3	10	20	1
4	1	3	4	10	20	0
4	2	3	5	20	30	1



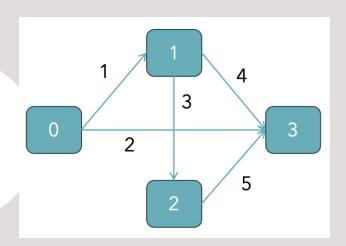
ID	1.time	1.status	2.time	2.status	3.time	3.status
5	10	1	20	0	20	1

ID	From	То	Trans	Start	Stop	Status
5	0	1	1	0	10	1
5	0	3	2	0	10	0
5	1	2	3	10	20	0
5	1	3	4	10	20	1

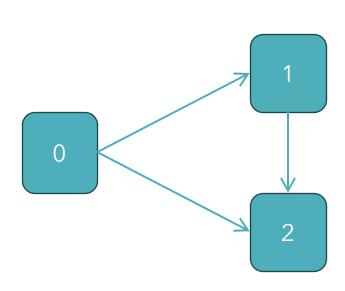


ID	1.time	1.status	2.time	2.status	3.time	3.status
6	20	0	20	0	20	1

ID	From	То	Trans	Start	Stop	Status
6	0	1	1	0	20	0
6	0	3	2	0	20	1



### Aalen-Johansen estimator



Irreversible illness-death model

$$\begin{split} \widehat{\mathbf{P}_{00}}(s,t) &= \prod_{s < u \le t} \left\{ 1 - \widehat{\lambda_{01}}(u) - \widehat{\lambda_{02}}(u) \right\} \\ \widehat{\mathbf{P}_{11}}(s,t) &= \prod_{s < u \le t} \left\{ 1 - \widehat{\lambda_{12}}(u) \right\} \\ \widehat{\mathbf{P}_{12}}(s,t) &= \sum_{s < u \le t} \widehat{\mathbf{P}_{11}}(s,u-)\widehat{\lambda_{12}}(u) \\ \widehat{\mathbf{P}_{01}}(s,t) &= \sum_{s < u \le t} \widehat{\mathbf{P}_{00}}(s,u-)\widehat{\lambda_{01}}(u)\widehat{\mathbf{P}_{11}}(u+,t) \\ \widehat{\mathbf{P}_{02}}(s,t) &= \sum_{s < u \le t} \widehat{\mathbf{P}_{00}}(s,u-)\widehat{\lambda_{01}}(u)\widehat{\mathbf{P}_{12}}(u+,t) \\ &+ \sum_{s < u \le t} \widehat{\mathbf{P}_{00}}(s,u-)\widehat{\lambda_{02}}(u) \end{split}$$

But if we would like to consider for the covariate pattern ...

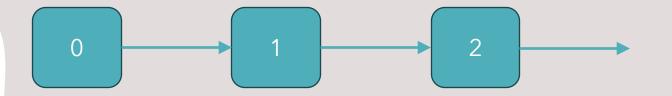
#### Covariate

Fixed at baseline

Transition specific

Time-varying within transition

Each transition has a different hazard function, which could have the same or different set of covariates



$$X = x_0$$

$$X = x_0 \longrightarrow X = x_1 \longrightarrow X = x_2 \longrightarrow$$

$$X = X_{0.1} \quad X_{0.2} \quad X_{0.3} \quad X_{1.1} \quad X_{1.2} \quad X_{2.1}$$

ID	From	То	Trans	Start	Stop	Status	HbA1c
3	0	1	1	0	10	1	7.2
3	0	3	2	0	10	0	7.2
3	1	2	3	10	20	1	7.2
3	1	3	4	10	20	0	7.2
3	2	3	5	20	30	0	7.2

ID	From	То	Trans	Start	Stop	Status	HbA1c
3	0	1	1	0	10	1	7.2
3	0	3	2	0	10	0	7
3	1	2	3	10	20	1	7.8
3	1	3	4	10	20	0	6.7
3	2	3	5	20	30	0	7

## Modeling

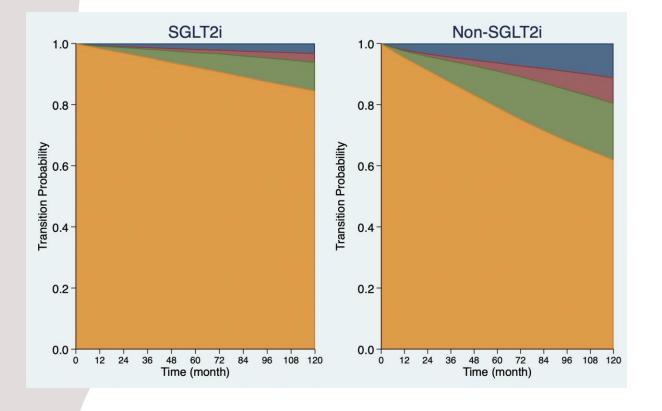
Cox PH model

Parametric survival model

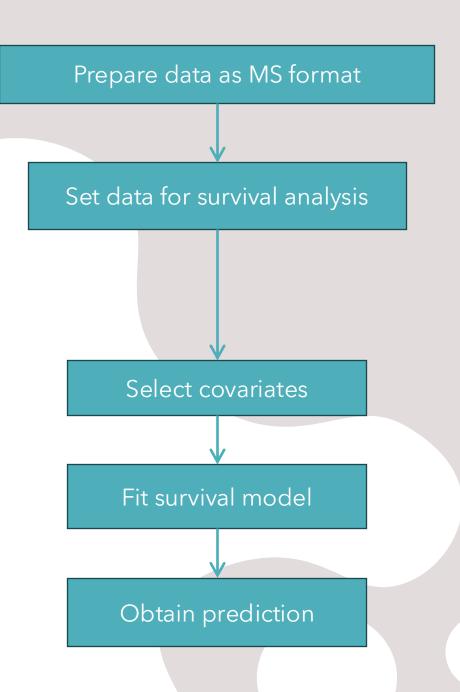
Spline-based model

## Transition probability

the probability of a subject being in a certain stage given the <u>subject's starting</u> <u>stage</u>, the time at which the subject began moving through the process, and the subject's covariate profile.



# Steps of analysis



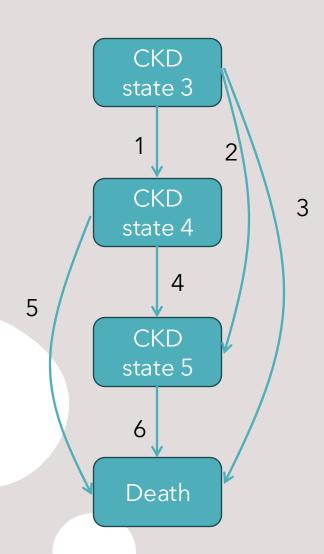
## Data for workshop

5266 patients

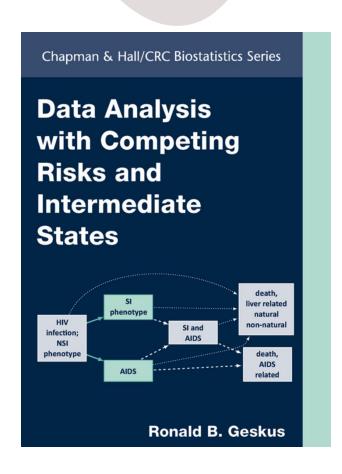
We want to know about the CKD progression

SGLT2i is the intervention of interest (compared with other 2<sup>nd</sup>-medications)

Match 1:2 by the date of prescription



#### STATA time!



<< Suggested reading