

# A Structural Description of Biases That Generate Immortal Time

Hernán MA et al. Epidemiology. 2025

Siriluck Manapunsopee M.Sc. Medical epidemiology 2023

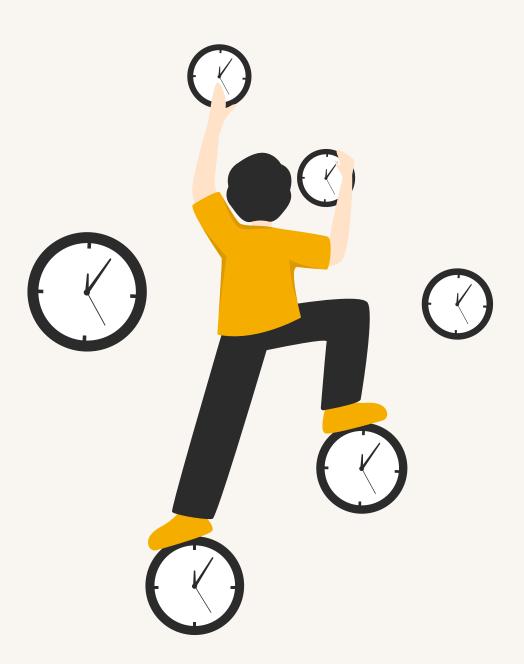
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### What is immortal time?

- Event -free time
- A period in survival analysis where individuals cannot experience the event of interest by definition
- If the event is death, the person is "immortal".
- Not truly immortal

# Why is it important?

• Survival analyses with immortal time yield biased effect estimates.



# Two situations that produce immortal time





### Due to selection

When an eligibility criteria is (incorrectly) applied after the start of follow-up.



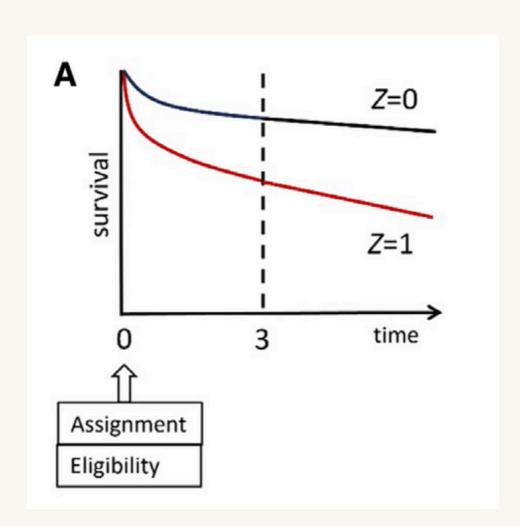
### Due to misclassification

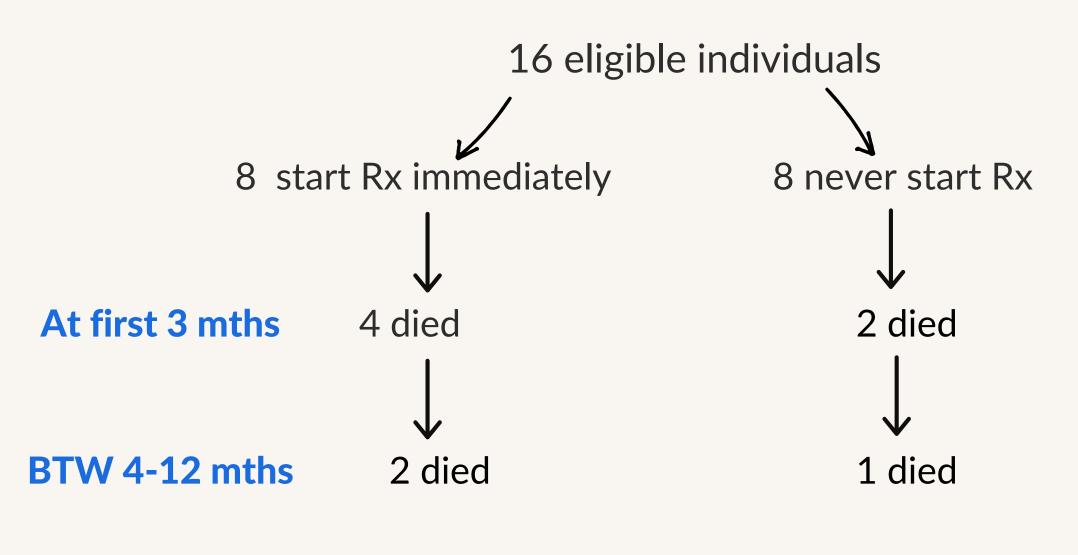
When misclassifying individuals into a treatment group that differs from the one they should have been assigned to.

# Immortal time due to Selection "Post-assignment Eligibility"



### Randomized trial





RCT: patients are assigned Rx only after eligibility has been met.

# Immortal time due to Selection "Post-assignment Eligibility"

# Large observational cohort: emulated target trial

individuals who met trial's eligibility

criteria

started Rx at time zero

no Rx at time zero

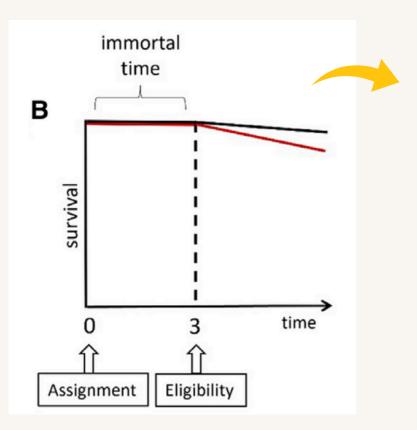
At first 3 mths

Record outcomes at 12 mths

Comparing the risks

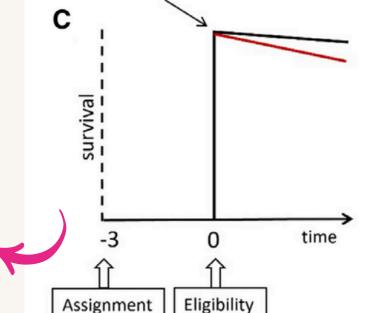


# Restrict the analysis to individuals who have completed 3 mths FU



When immortal time is included in calculation: "Immortal time bias".

Enroll individuals who have started Rx before and analysis start at the time of Rx initiation



"landmark"

: "Prevalent user bias"

Not include immortal time in calculation

Differential depletion of susceptibles

# Immortal time due to Selection "Post-assignment Eligibility"

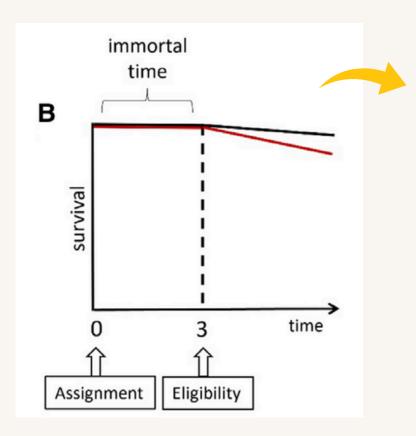
Large observational cohort: emulated target trial

### What to do

- All eligibility criteria are defined at time zero so no selection occurs after treatment assignment.
- Explicitly emulating target trial based on data for all eligible individuals from the time of treatment assignment

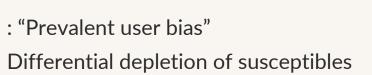


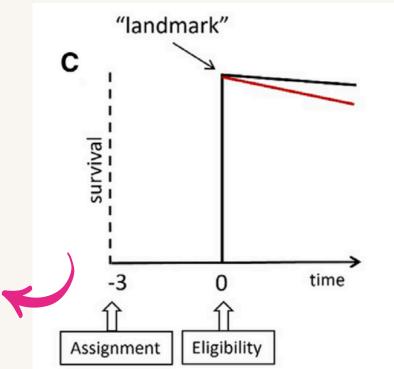
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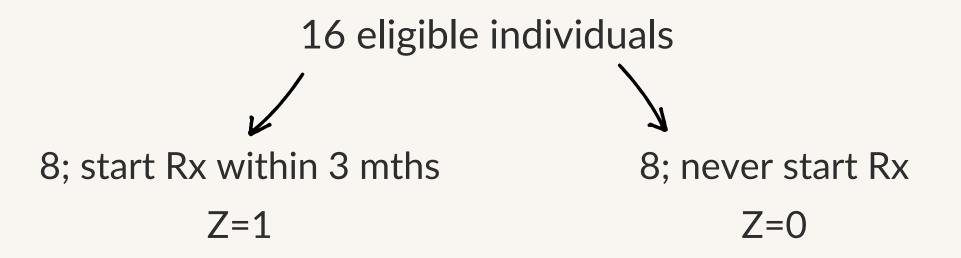


### Immortal time due to Misclassification

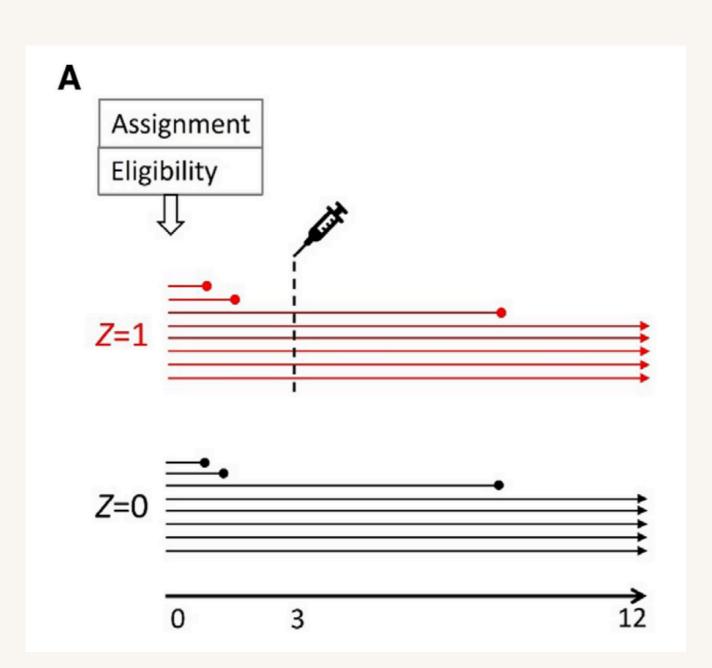


When some strategies under study cannot be distinguished at time zero

New randomized trial with full adherence

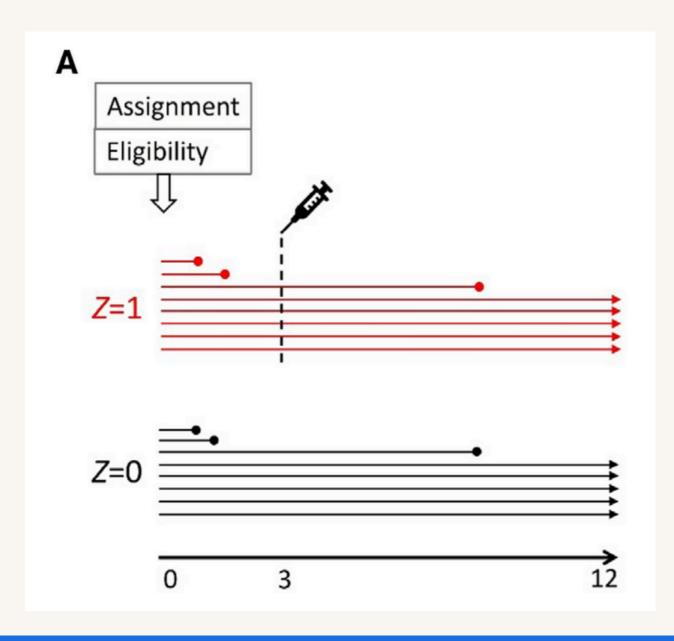


Z=1 has a 3 month 'grace period' within which treatment should be started.

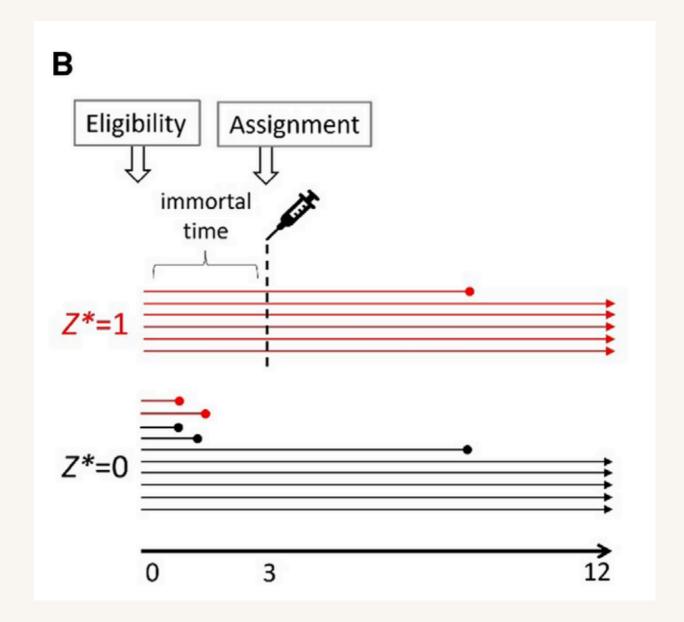


### Immortal time due to Misclassification

Observational study include information on treatment Am and outcome Ym overtime



The patients who died during grace period were classifed to  $Z^* = 0$ .



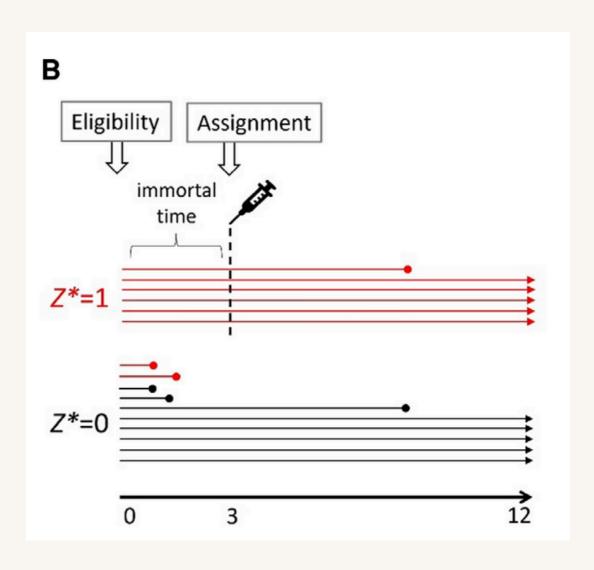
What if, assign to Z\*=1 if the individual received treatment within 3 months (i.e., Am=1 for m (month)=0,1, or 2)

"bias in assignment of subjects"

### Immortal time due to Misclassification



This form of misclassification of assignment is relevant for any strategies that are not distinguishable at time zero (e.g., strategies with grace or waiting periods, strategies of different durations of treatment).



### What to do

Several approaches to prevent the emergence of this type of immortal time.

- 1. Changing the causal question: strategies that are distinguishable at time zero
- 2. Cloning followed by censoring
- 3. g-formula

# Cloning followed by censoring



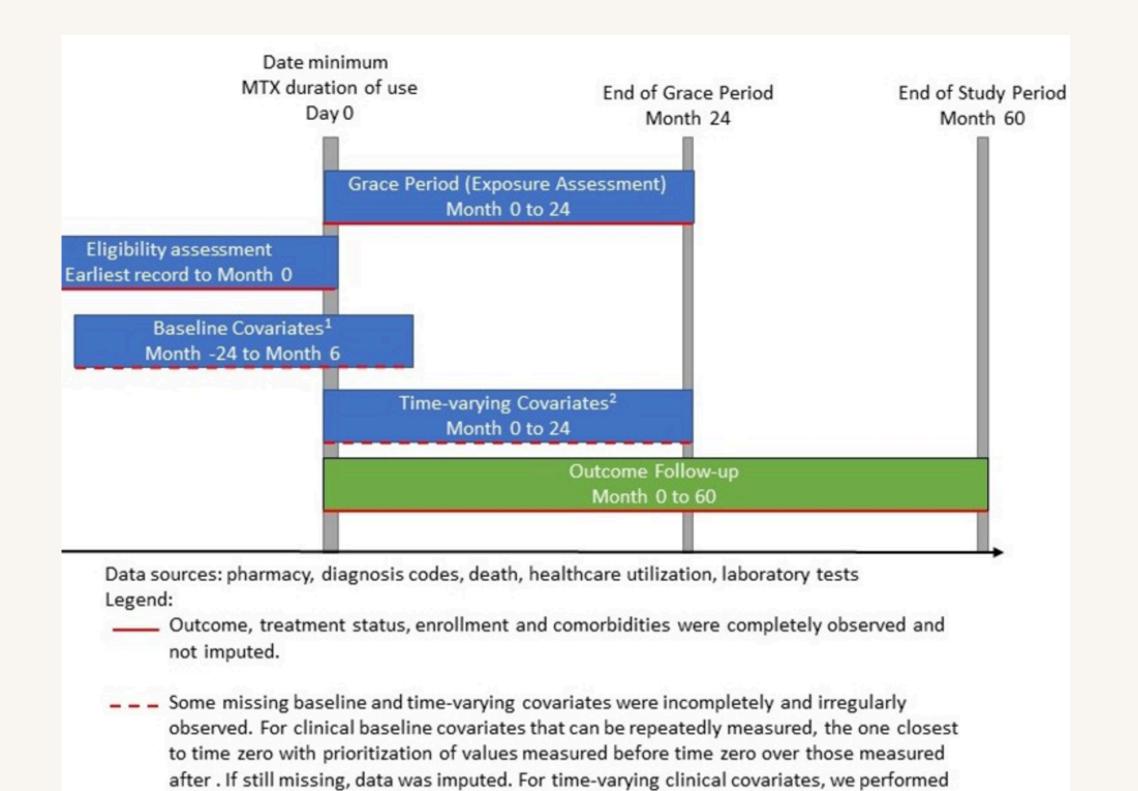
At time zero, some individuals have data compatible with more than one strategy.

We can create as many "clones" of each individual as treatment strategies the individual's data are compatible with at baseline.

Each clone is assigned to a single strategy and censored if and when the clone deviates from it (i.e., stops having data compatible with it).

Adjustment via inverse-probability weighting is required.

Computation of the variance for cloning estimators is accounted.



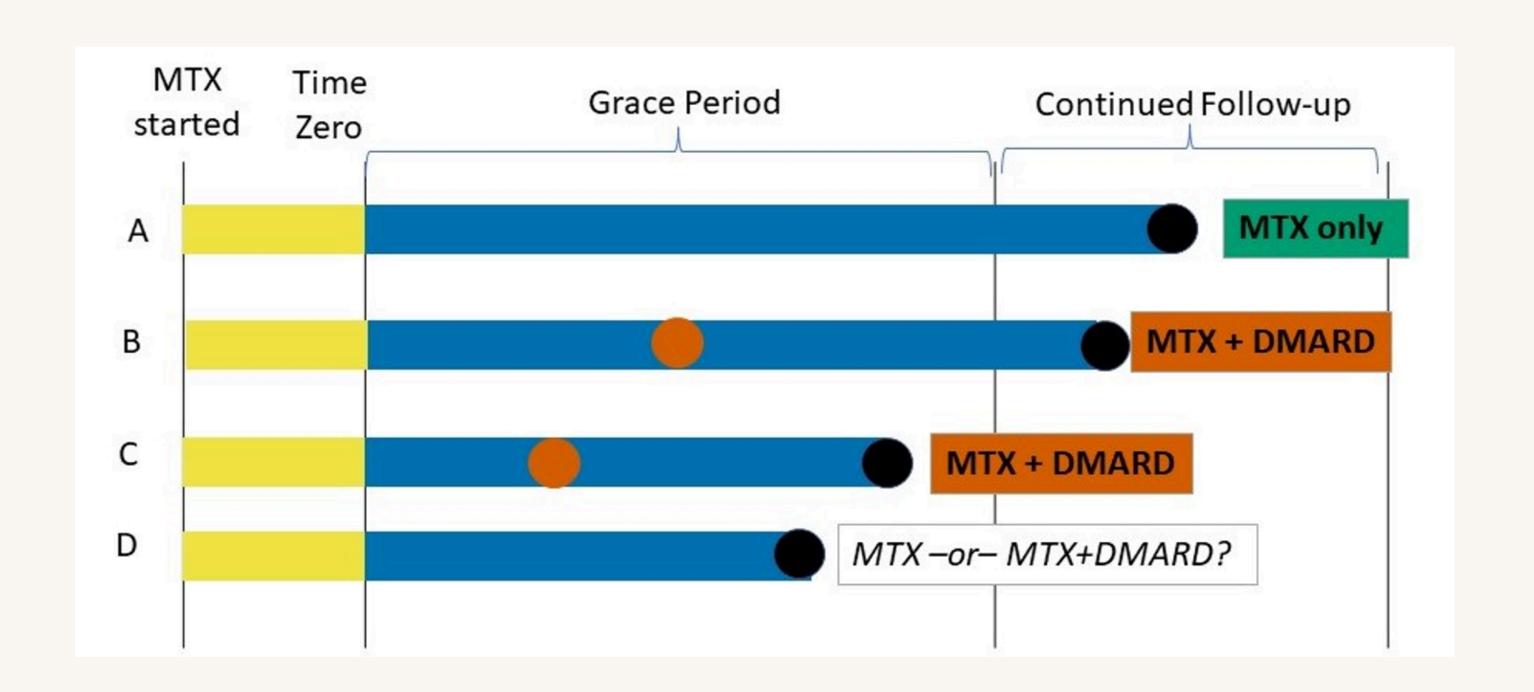
Note that missing data from possible out-of-system diagnosis or healthcare visits is possible for all

last-observation carried forward.

variables.



## Misclassification





# Thank you