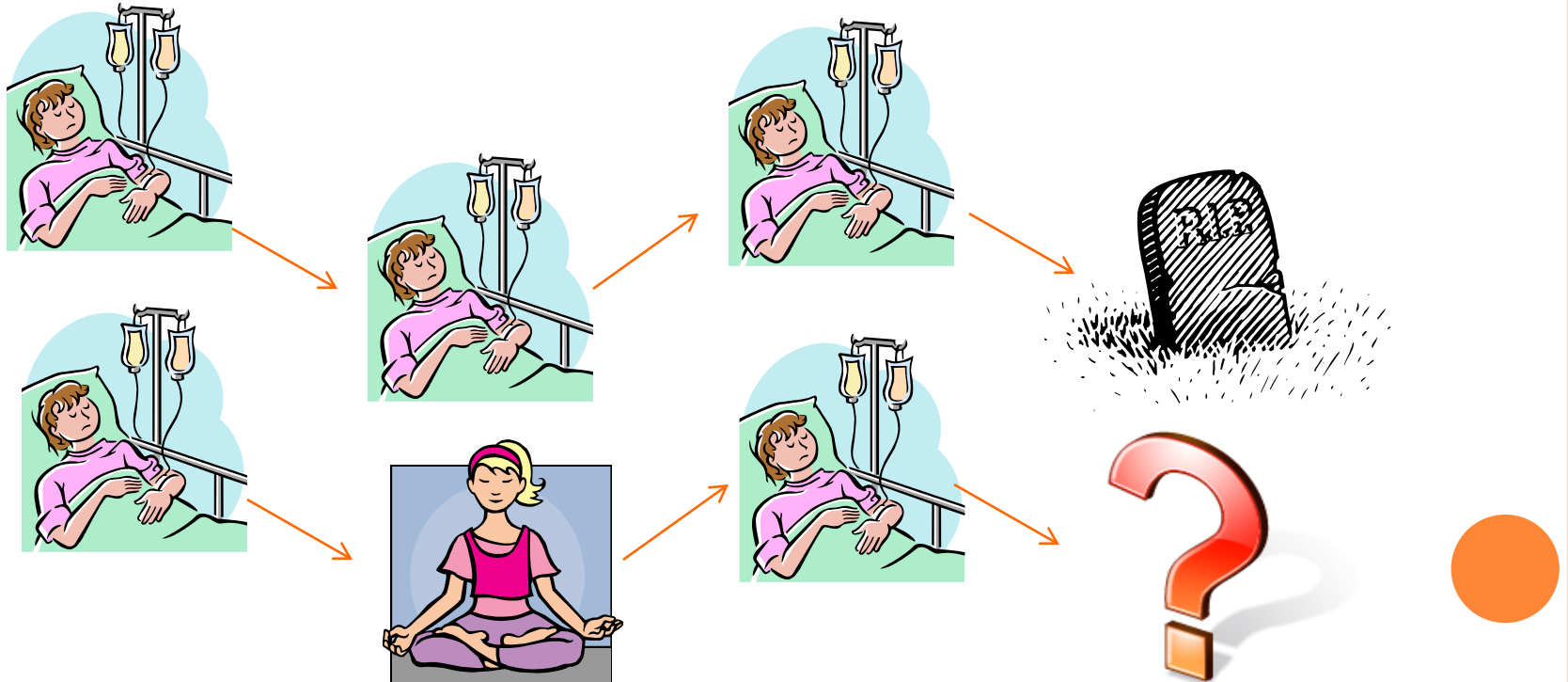


MARKOV DECISION TREE

Presenter: Jiaru Bai

MARKOV CHAIN

- Not all problems can be portrayed in a standard decision tree, most medical problems are cyclical and the condition of patients can change over time. A Markov process is characterized by recurrent states over time.



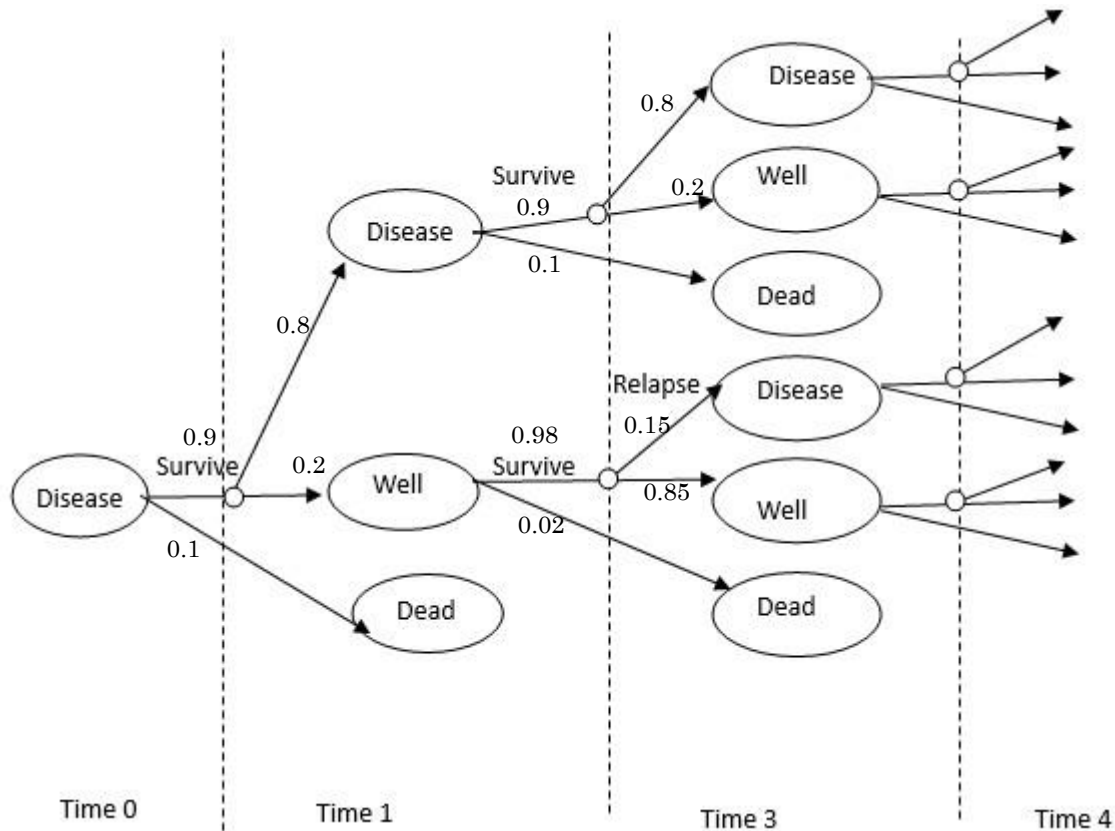
What about in five years?

AN EXAMPLE OF MARKOV PROCESS

- Consider this Markov process (You can find this example in the Treeage software package: Three-State Markov):
 - Here's a kind of disease, patients who get this disease can recover but could result in a relapse.
 - In each cycle or period (let's say a year), 10% of the patients who catch this disease will die immediately.
 - Among the rest of them who survive at first, after treatment, only 20% of them will recover and the rest will stay sick;
 - Even if people recovered, in one year, there is still 2% chance that they could die. And if they stay alive, 15% of them will catch this disease again.
- States: disease, well, dead.
- Question: How many people (proportion of the population) will be alive in 50 years?



PUT THE PROBLEM IN A DECISION TREE



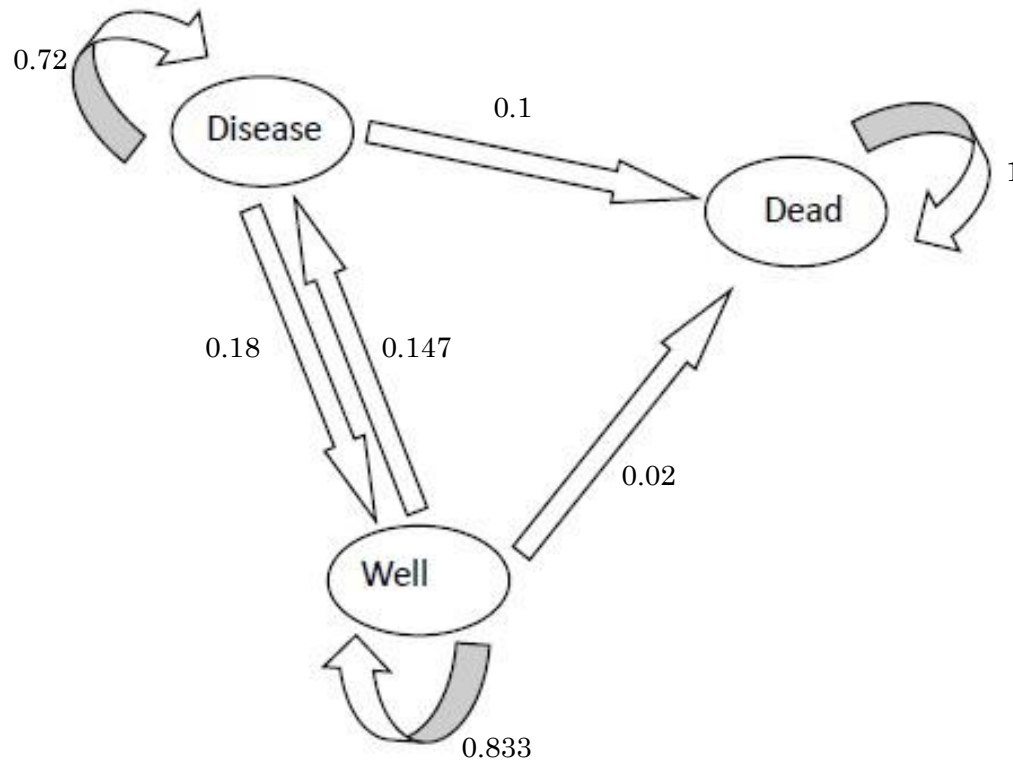
It becomes more complicated when time passes by.
There are some substructures repeating in the tree.



ANOTHER INTERPRETATION

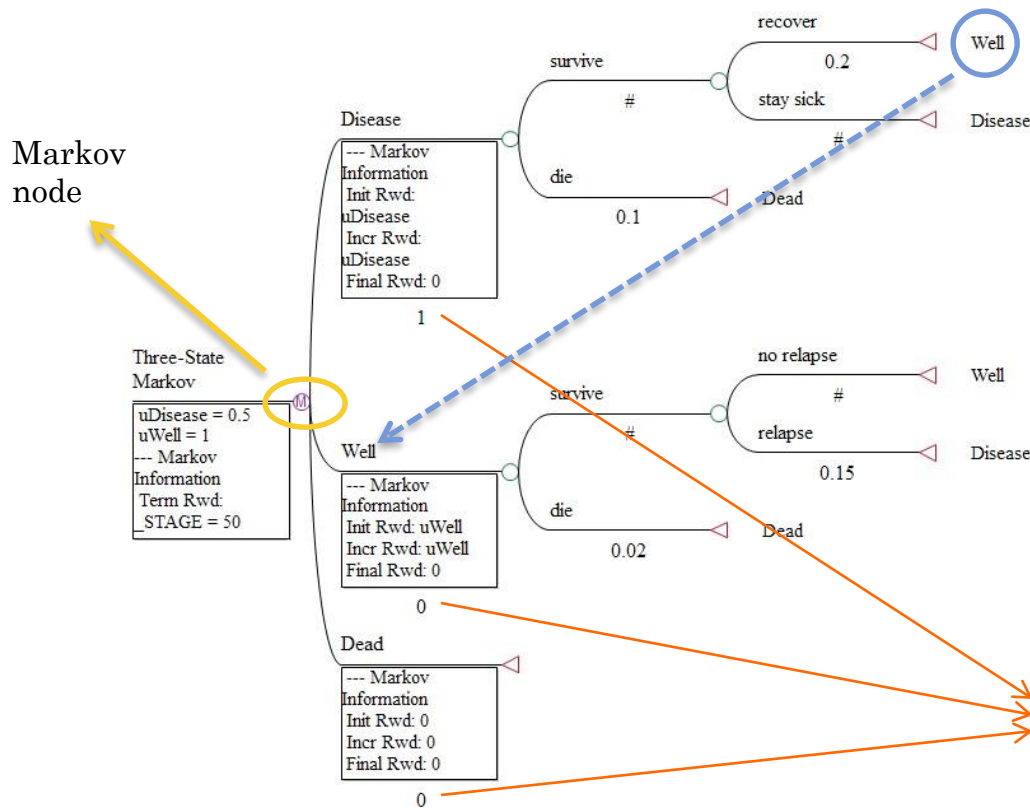
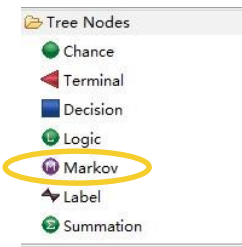
- **Absorbing** states. Being dead is an absorbing state.

B. Usual way a Markov process is drawn with nodes and arcs



A MARKOV TREE

- A Markov tree can be used to present a Markov process.
- A Markov tree is often composed of the following parts: structure, probabilities, rewards (quality adjusted life years) and termination condition.



Note: We changed uDisease from 1 to 0.5.

ANALYSIS OF A MARKOV TREE

- **Cohort (expected value) analysis:**
 - Compute the probability distribution after n stages.
 - Compute reward (quality) and cumulative reward after n stages.

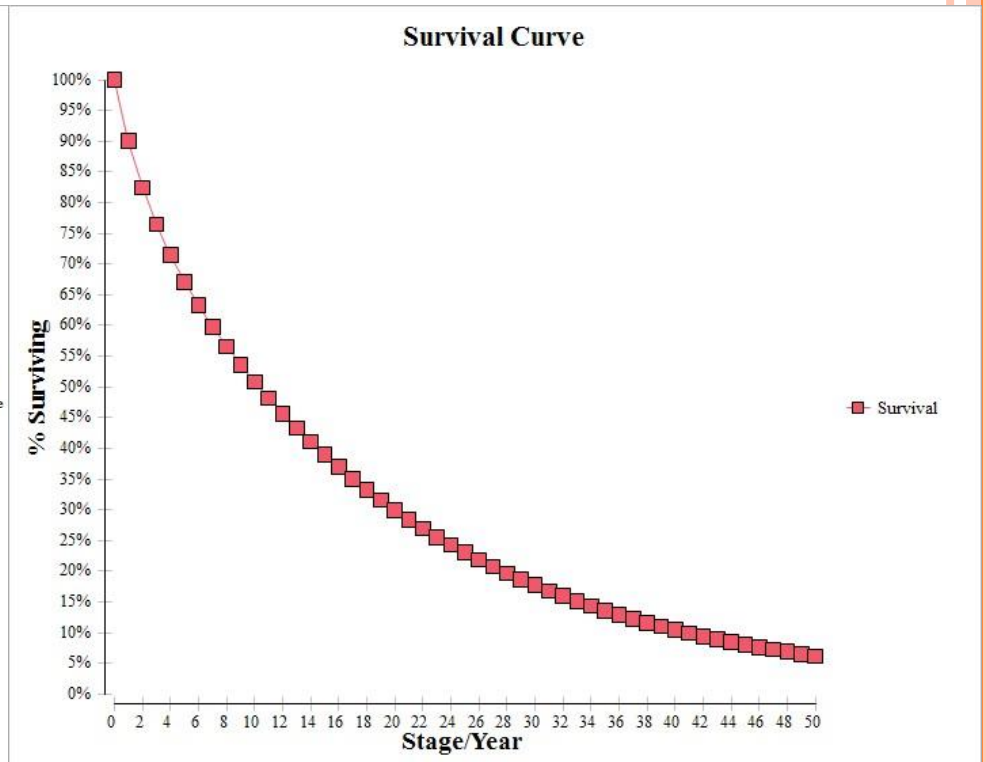
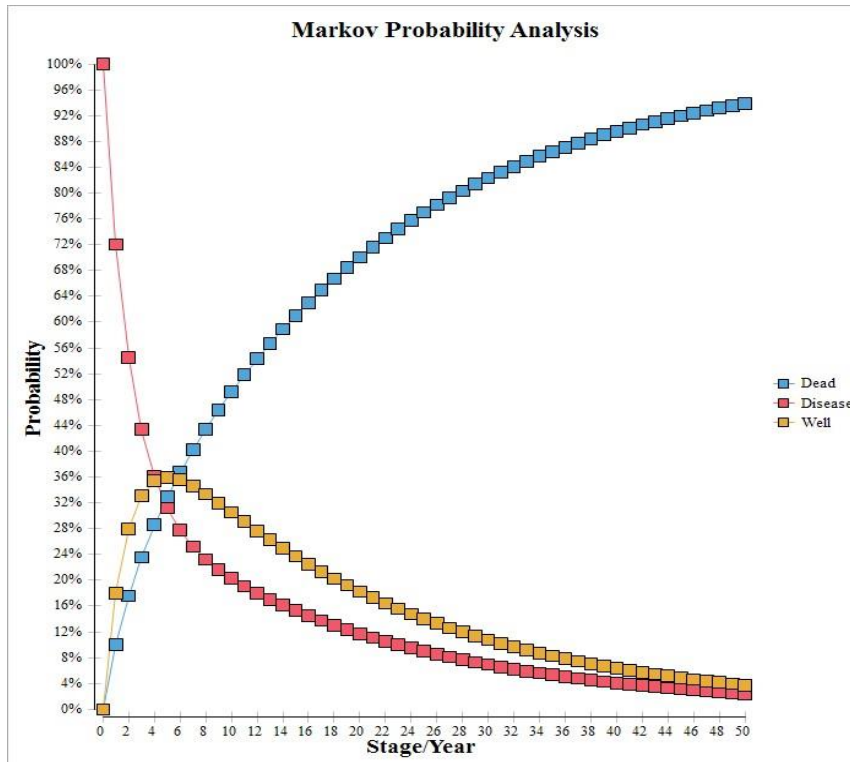
Markov Cohort (Quick)

Stage	State	Probability	State Reward	Stage Reward	Cumulative Reward	Actions
▲ Stage 0				0.5	0.5	— Text Reports —
	Disease	1	0.5			Summary Report
	Well	0	0			— Charts —
	Dead	0	0			State Prob.
▲ Stage 1				0.54	1.04	Survival Curve
	Disease	0.72	0.36			State Reward
	Well	0.18	0.18			Stage Reward
	Dead	0.1	0			Cumulative Reward
▲ Stage 2				0.55197	1.59197	
	Disease	0.54486	0.27243			
	Well	0.27954	0.27954			
	Dead	0.1756	0			
▲ Stage 3				0.54763	2.1396	
	Disease	0.43339	0.2167			
	Well	0.33093	0.33093			
	Dead	0.23568	0			

ANALYSIS OF A MARKOV TREE

Markov Cohort Graphical Output:

- State probabilities
- Survival curve
- State reward
- Stage reward
- Cumulative reward



MARKOV DECISION TREE

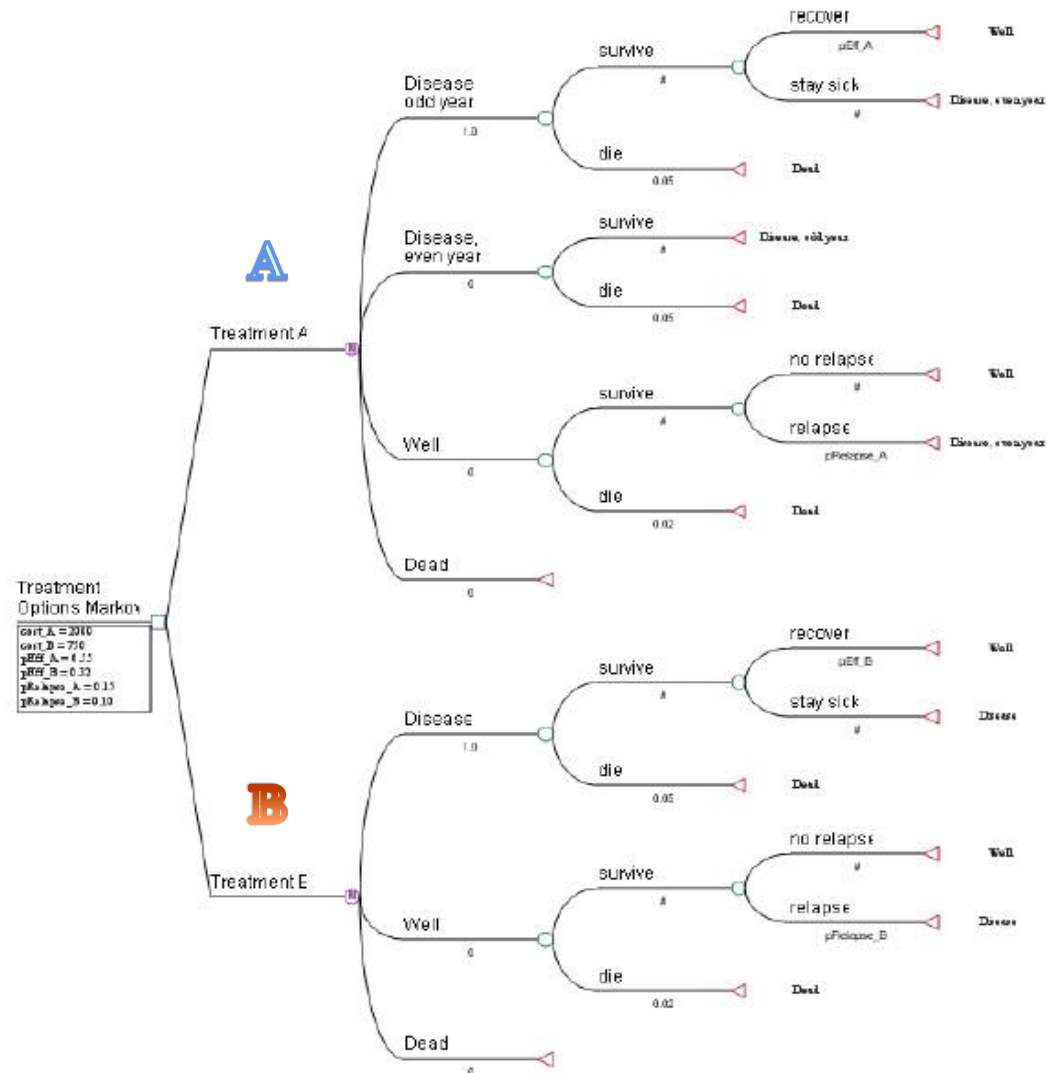
If you are interested in comparing two treatments/
Markov Trees:

- Markov cycle trees can be appended to paths in a decision tree anywhere you might place a terminal node.
- A decision tree needs you to compare two choices/treatments.
- A Markov decision tree includes Markov subtrees.

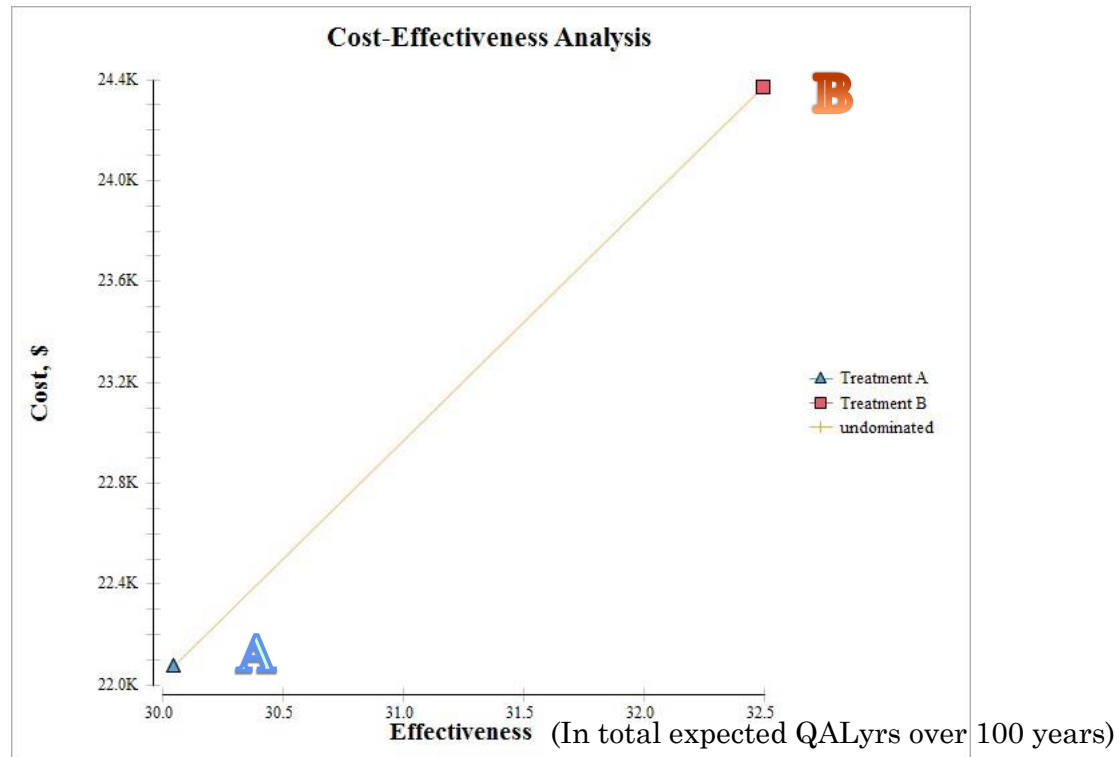


A MARKOV DECISION TREE

- Treatment A is presumed to be faster acting, but cannot be used long term due to higher relapse rate.
- Treatment B is slower acting, but can be used on a maintenance basis over a prolonged period, effectively preventing more relapses.



COST-EFFECTIVENESS



- Treatment A has lower cost but lower effectiveness, while treatment B has higher cost but higher effectiveness.



* EXTENSIONS

- The default setting of a Markov decision tree is that the transition probabilities are the same for each stage (probability of going from well to sick is the same each stage). But you can introduce a tracker and tunnel variable to make it history-dependent (e.g. different transition probabilities for surgery 1, 2 and 3).



REFERENCES AND USEFUL LINKS

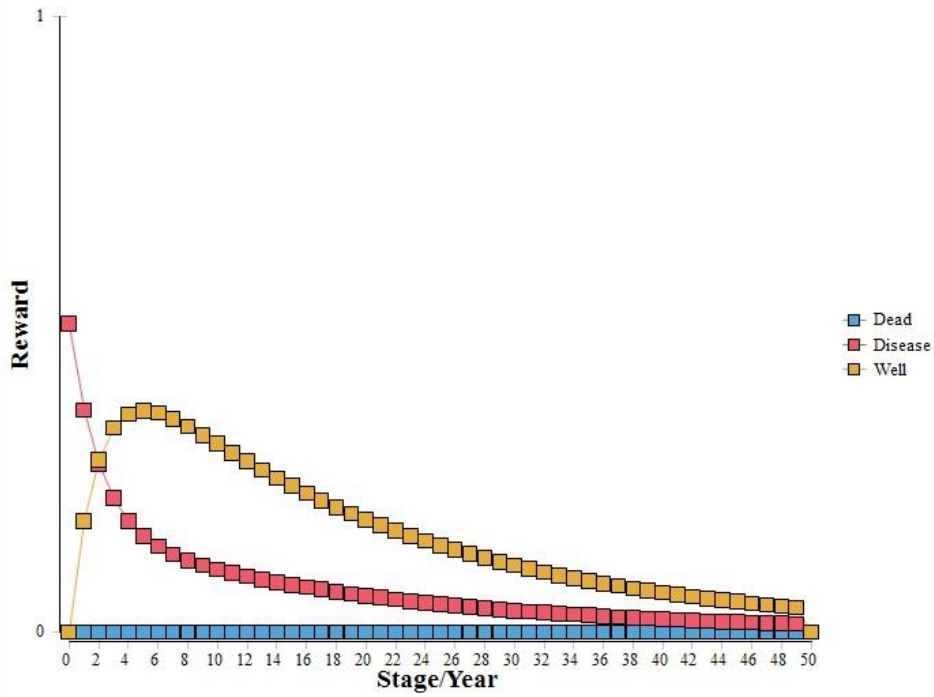
- TreeAge Pro 2013 User's Manual
- File to download: How to build a Markov tree(TreeAge Software)
- File to download: Markov Decision Tree
- Also you can find corresponding chapter in manual, page 434.



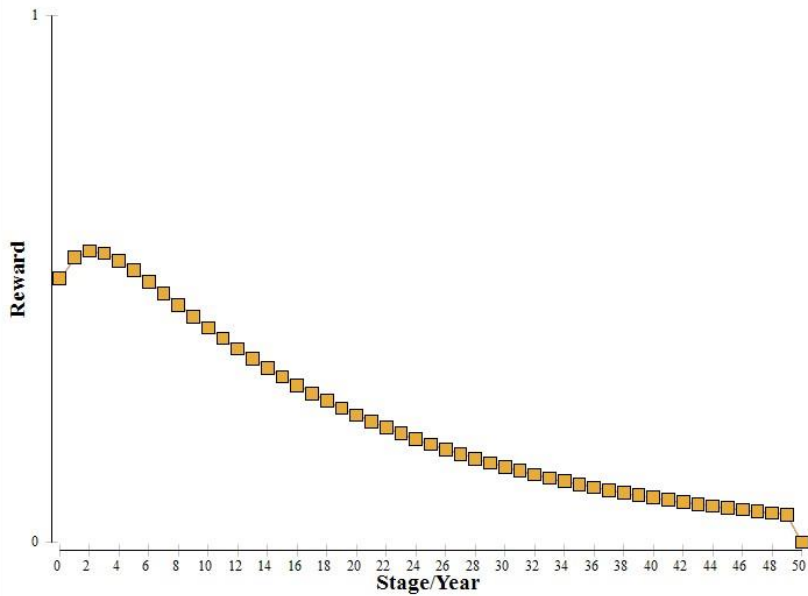
○ Thank you!



Markov Reward Per State



Markov Reward Per Stage



Markov Cumulative Reward

