

Tegaderm CHG IV Securement Dressing for Central Venous and Arterial Catheter Insertion Sites

A decision tree example with probabilistic sensitivity analysis

Andrew J. Sims

2021-02-11

Introduction

This vignette is an example of modelling a decision tree using the `rdecision` package, with probabilistic sensitivity analysis. It is based on the model reported by Jenks *et al* [1] in which a transparent dressing used to secure vascular catheters (Tegaderm CHG) was compared with a standard dressing.

Model variables

Thirteen variables were used in the model. The choice of variables, their distributions and their parameters are taken from table 3 of Jenks *et al* [1], with the following additional information:

- For variables with lognormal uncertainty, the manufacturer synthesized a log normal distribution as $\exp(\alpha + \beta r())$ where $r()$ is a random draw from a standard normal distribution. This does not follow the parametrizations of the `LogNormModVar` provided in `rdecision`, but can be reproduced using expression model variables. Three standard normal model variables were introduced for this purpose.
- The values for α and β for the hazard ratio for CRBSI and LSI were -0.911 and -0.393 respectively, and 1.482 and 0.490 for the relative risk of dermatitis.
- The probabilities of CRBSI and LSI for standard dressings (p) were modified by the hazard ratio for Tegaderm using the form $(1 - (1 - p)^h)$ where h is the hazard ratio. Relative risks were applied as multipliers.
- The point estimate cost of CRBSI was £9900, not £9990, although the parameters (198,50) are quoted correctly.

The model variables were constructed as follows:

```
# standard normals
n1 <- NormModVar$new("SN1","", 0, 1)
n2 <- NormModVar$new("SN2","", 0, 1)
n3 <- NormModVar$new("SN3","", 0, 1)

# clinical variables
r.CRBSI <- NormModVar$new(
  'Baseline CRBSI rate', '/1000 catheter days', mu=1.48, sigma=0.074
)
hr.CRBSI <- ExprModVar$new("Tegaderm CRBSI HR", "HR", rlang::quo(exp(-0.911-0.393*n1)))
hr.LSI <- ExprModVar$new("Tegaderm LSI HR", "HR", rlang::quo(exp(-0.911-0.393*n2)))
r.Dermatitis <- NormModVar$new(
  'Baseline dermatitis risk', '/catheter', mu=0.0026, sigma=0.00026
```

```

)
rr.Dermatitis <- ExprModVar$new("Tegaderm LSI HR", "HR", rlang::quo(exp(1.482-0.490*n3)))

# cost variables
c.CRBSI <- GammaModVar$new(
  'CRBSI cost', 'GBP', shape=198.0, scale=50
)
c.Dermatitis <- GammaModVar$new(
  'Dermatitis cost', 'GBP', shape=30, scale=5
)
c.LSI <- GammaModVar$new(
  'LSI cost', 'GBP', shape=50, scale=5
)
n.catheters <- NormModVar$new(
  'No. catheters', 'catheters', mu=3, sigma=0.3
)
c.Tegaderm <- ExprModVar$new(
  "Tegaderm CHG cost", "GBP", rlang::quo(6.21*n.catheters)
)
c.Standard <- ExprModVar$new(
  "Standard dressing cost", "GBP", rlang::quo(1.34*n.catheters)
)
n.cathdays <- NormModVar$new(
  'No. days with catheter', 'days', mu=10, sigma=2
)

```

Model variable expressions

Variables in the model may be included in the decision tree via mathematical expressions, which involve model variables and are themselves model variables. Forms of expression involving R's numerical functions and multiple model variables are supported, provided they conform to R syntax. The following code creates the model variable expressions to be used as values in the decision tree edges. For probabilities, the convention $q = 1 - p$ is used to ensure that the sum of probabilities leaving each chance node sums to one.

```

# probabilities
p.Dermatitis.S <- ExprModVar$new(
  'P(dermatitis|standard dressing)', 'P',
  rlang::quo(r.Dermatitis*n.catheters)
)
q.Dermatitis.S <- ExprModVar$new(
  'Q(dermatitis|standard dressing)', '1-P',
  rlang::quo(1-p.Dermatitis.S)
)
p.Dermatitis.T <- ExprModVar$new(
  'P(dermatitis|Tegaderm)', 'P',
  rlang::quo(r.Dermatitis*rr.Dermatitis*n.catheters)
)
q.Dermatitis.T <- ExprModVar$new(
  'Q(dermatitis|Tegaderm)', '1-P',
  rlang::quo(1-p.Dermatitis.T)
)

```

```

p.LSI.S <- NormModVar$new(
  'P(LSI|Standard)', '/patient', mu=0.1, sigma=0.01
)
q.LSI.S <- ExprModVar$new(
  'Q(LSI|Standard)', '1-P', rlang::quo(1-p.LSI.S)
)
p.LSI.T <- ExprModVar$new(
  'P(LSI|Tegaderm)', 'P', rlang::quo(1-(1-p.LSI.S)^hr.LSI)
)
q.LSI.T <- ExprModVar$new(
  'Q(LSI|Tegaderm)', '1-P', rlang::quo(1-p.LSI.T)
)

p.CRBSI.S <- ExprModVar$new(
  'P(CRBSI|standard dressing)', 'P', rlang::quo(r.CRBSI*n.cathdays/1000)
)
q.CRBSI.S <- ExprModVar$new(
  'Q(CRBSI|standard dressing)', '1-P', rlang::quo(1-p.CRBSI.S)
)
p.CRBSI.T <- ExprModVar$new(
  'P(CRBSI|Tegaderm)', 'P', rlang::quo(1-(1-r.CRBSI*n.cathdays/1000)^hr.CRBSI)
)
q.CRBSI.T <- ExprModVar$new(
  'Q(CRBSI|Tegaderm)', '1-P', rlang::quo(1-p.CRBSI.T)
)

```

The decision tree

The following code constructs the decision tree based on figure 2 of Jenks *et al* [1]. In the formulation used by `rdecision`, the decision tree is constructed from sets of decision, chance and leaf nodes and from edges (actions and reactions). Leaf nodes are synonymous with pathways in Briggs' terminology [2]. The time horizon is not stated explicitly in the model, and is assumed to be 7 days. It was implied that the time horizon was ICU stay plus some follow-up, and the costs reflect those incurred in that period, so the assumption of 7 days does not affect the `rdecision` implementation of the model.

The tree is somewhat more complex than figure 2 of Jenks *et al* because it allows for patients to have more than one adverse event (AE) during their stay (whereas figure 2 implies that only one event per patient is possible). The rates of AE were estimated independently, and allow for multiple events.

```

# create decision tree
th <- as.difftime(7, units="days")
# standard dressing
t01 <- LeafNode$new("t01", interval=th)
t02 <- LeafNode$new("t02", interval=th)
c01 <- ChanceNode$new()
e01 <- Reaction$new(c01,t01,p=p.Dermatitis.S,cost=c.Dermatitis)
e02 <- Reaction$new(c01,t02,p=q.Dermatitis.S,cost=0)
#
t03 <- LeafNode$new("t03", interval=th)
t04 <- LeafNode$new("t04", interval=th)
c02 <- ChanceNode$new()
e03 <- Reaction$new(c02,t03,p=p.Dermatitis.S,cost=c.Dermatitis)
e04 <- Reaction$new(c02,t04,p=q.Dermatitis.S,cost=0)

```

```

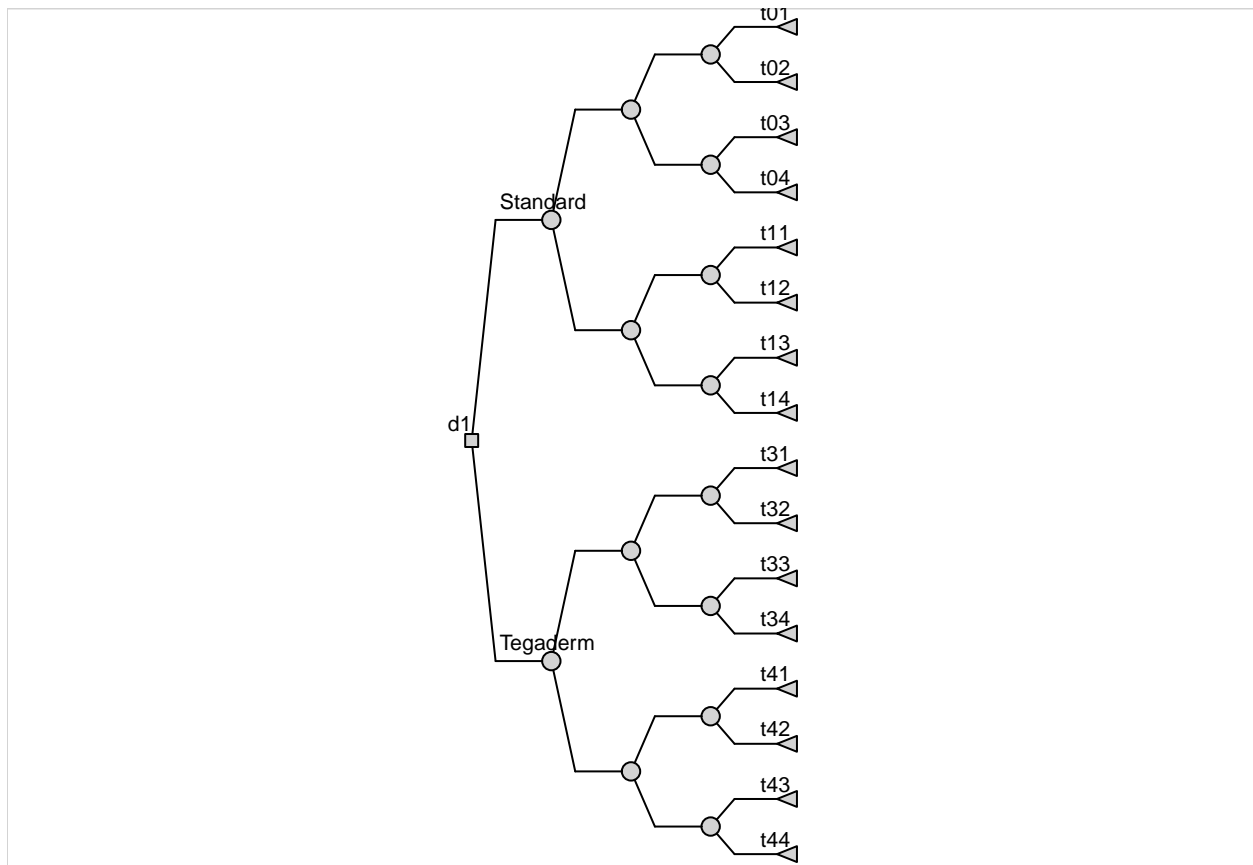
#
c03 <- ChanceNode$new()
e05 <- Reaction$new(c03,c01,p=p.LSI.S,cost=c.LSI)
e06 <- Reaction$new(c03,c02,p=q.LSI.S,cost=0)
#
t11 <- LeafNode$new("t11", interval=th)
t12 <- LeafNode$new("t12", interval=th)
c11 <- ChanceNode$new()
e11 <- Reaction$new(c11,t11,p=p.Dermatitis.S,cost=c.Dermatitis)
e12 <- Reaction$new(c11,t12,p=q.Dermatitis.S,cost=0)
#
t13 <- LeafNode$new("t13", interval=th)
t14 <- LeafNode$new("t14", interval=th)
c12 <- ChanceNode$new()
e13 <- Reaction$new(c12,t13,p=p.Dermatitis.S,cost=c.Dermatitis)
e14 <- Reaction$new(c12,t14,p=q.Dermatitis.S,cost=0)
#
c13 <- ChanceNode$new()
e15 <- Reaction$new(c13,c11,p=p.LSI.S,cost=c.LSI)
e16 <- Reaction$new(c13,c12,p=q.LSI.S,cost=0)
#
c23 <- ChanceNode$new()
e21 <- Reaction$new(c23,c03,p=p.CRBSI.S,cost=c.CRBSI)
e22 <- Reaction$new(c23,c13,p=q.CRBSI.S,cost=0)
#
# Tegaderm branch
t31 <- LeafNode$new("t31", interval=th)
t32 <- LeafNode$new("t32", interval=th)
c31 <- ChanceNode$new()
e31 <- Reaction$new(c31,t31,p=p.Dermatitis.T,cost=c.Dermatitis)
e32 <- Reaction$new(c31,t32,p=q.Dermatitis.T,cost=0)
#
t33 <- LeafNode$new("t33", interval=th)
t34 <- LeafNode$new("t34", interval=th)
c32 <- ChanceNode$new()
e33 <- Reaction$new(c32,t33,p=p.Dermatitis.T,cost=c.Dermatitis)
e34 <- Reaction$new(c32,t34,p=q.Dermatitis.T,cost=0)
#
c33 <- ChanceNode$new()
e35 <- Reaction$new(c33,c31,p=p.LSI.T,cost=c.LSI)
e36 <- Reaction$new(c33,c32,p=q.LSI.T,cost=0)
#
t41 <- LeafNode$new("t41", interval=th)
t42 <- LeafNode$new("t42", interval=th)
c41 <- ChanceNode$new()
e41 <- Reaction$new(c41,t41,p=p.Dermatitis.T,cost=c.Dermatitis)
e42 <- Reaction$new(c41,t42,p=q.Dermatitis.T,cost=0)
#
t43 <- LeafNode$new("t43", interval=th)
t44 <- LeafNode$new("t44", interval=th)
c42 <- ChanceNode$new()
e43 <- Reaction$new(c42,t43,p=p.Dermatitis.T,cost=c.Dermatitis)
e44 <- Reaction$new(c42,t44,p=q.Dermatitis.T,cost=0)

```

```

#
c43 <- ChanceNode$new()
e45 <- Reaction$new(c43,c41,p=p.LSI.T,cost=c.LSI)
e46 <- Reaction$new(c43,c42,p=q.LSI.T,cost=0)
#
c53 <- ChanceNode$new()
e51 <- Reaction$new(c53,c43,p=p.CRBSI.T,cost=c.CRBSI)
e52 <- Reaction$new(c53,c33,p=q.CRBSI.T,cost=0)
#
# decision node
d1 <- DecisionNode$new("d1")
e9 <- Action$new(d1,c23,label="Standard",cost=c.Standard)
e10 <- Action$new(d1,c53,label="Tegaderm",cost=c.Tegaderm)
#
# create decision tree
V <- list(d1,c01,c02,c03,c11,c12,c13,c23,c31,c32,c33,c41,c42,c43,c53,
          t01,t02,t03,t04,t11,t12,t13,t14,t31,t32,t33,t34,t41,t42,t43,t44)
E <- list(e01,e02,e03,e04,e05,e06,e11,e12,e13,e14,e15,e16,e21,e22,
          e31,e32,e33,e34,e35,e36,e41,e42,e43,e44,e45,e46,e51,e52,e9,e10)
DT <- DecisionTree$new(V,E)

```



In the manufacturer's model, the uncertainties in the probabilities associated with the polytomous chance nodes were modelled as independent variables. This is not recommended because there is a chance that a particular run of the PSA will yield probabilities that are outside the range $[0,1]$. Representing the uncertain probabilities with draws from a Dirichlet distribution is preferred.

Summary of the model

The model variables associated with actions, reactions and leaf nodes can be tabulated using the method `modvar_table`. This returns a data frame describing each variable, its description, units and uncertainty distribution. Variables inheriting from type `ModVar` will be included in the tabulation; regular numeric values will not be listed. In the Tegaderm model, the complete structure is as follows:

Description	Distribution
Dermatitis cost	Ga(30,5)
P(dermatitis standard dressing)	r.Dermatitis * n.catheters
Baseline dermatitis risk	N(0.0026,0.00026)
No. catheters	N(3,0.3)
Q(dermatitis standard dressing)	1 - p.Dermatitis.S
LSI cost	Ga(50,5)
P(LSI Standard)	N(0.1,0.01)
Q(LSI Standard)	1 - p.LSI.S
CRBSI cost	Ga(198,50)
P(CRBSI standard dressing)	r.CRBSI * n.cathdays/1000
Baseline CRBSI rate	N(1.48,0.074)
No. days with catheter	N(10,2)
Q(CRBSI standard dressing)	1 - p.CRBSI.S
P(dermatitis Tegaderm)	r.Dermatitis * rr.Dermatitis * n.catheters
Tegaderm LSI HR	exp(1.482 - 0.49 * n3)
SN3	N(0,1)
Q(dermatitis Tegaderm)	1 - p.Dermatitis.T
P(LSI Tegaderm)	1 - (1 - p.LSI.S)^hr.LSI
Tegaderm LSI HR	exp(-0.911 - 0.393 * n2)
SN2	N(0,1)
Q(LSI Tegaderm)	1 - p.LSI.T
P(CRBSI Tegaderm)	1 - (1 - r.CRBSI * n.cathdays/1000)^hr.CRBSI
Tegaderm CRBSI HR	exp(-0.911 - 0.393 * n1)
SN1	N(0,1)
Q(CRBSI Tegaderm)	1 - p.CRBSI.T
Standard dressing cost	1.34 * n.catheters
Tegaderm CHG cost	6.21 * n.catheters

Point estimates and distributions of model variables

The point estimates, units and distributional properties are obtained from the same call, in the remaining columns.

Description	Units	Mean	Q2.5	Q97.5
Dermatitis cost	GBP	150	101	208
P(dermatitis standard dressing)	P	0.0078	0.00564	0.00999
Baseline dermatitis risk	/catheter	0.0026	0.00209	0.00311
No. catheters	catheters	3	2.41	3.59
Q(dermatitis standard dressing)	1-P	0.992	0.99	0.994
LSI cost	GBP	250	186	324
P(LSI Standard)	/patient	0.1	0.0804	0.12
Q(LSI Standard)	1-P	0.9	0.88	0.92
CRBSI cost	GBP	9900	8569	11326

Description	Units	Mean	Q2.5	Q97.5
P(CRBSI standard dressing)	P	0.0148	0.00873	0.0209
Baseline CRBSI rate	/1000 catheter days	1.48	1.33	1.63
No. days with catheter	days	10	6.08	13.9
Q(CRBSI standard dressing)	1-P	0.985	0.979	0.991
P(dermatitis Tegaderm)	P	0.0343	0.0124	0.0944
Tegaderm LSI HR	HR	4.4	1.69	11.9
SN3		0	-1.96	1.96
Q(dermatitis Tegaderm)	1-P	0.966	0.906	0.987
P(LSI Tegaderm)	P	0.0415	0.0191	0.088
Tegaderm LSI HR	HR	0.402	0.187	0.845
SN2		0	-1.96	1.96
Q(LSI Tegaderm)	1-P	0.959	0.913	0.982
P(CRBSI Tegaderm)	P	0.00598	0.00251	0.0142
Tegaderm CRBSI HR	HR	0.402	0.185	0.848
SN1		0	-1.96	1.96
Q(CRBSI Tegaderm)	1-P	0.994	0.987	0.998
Standard dressing cost	GBP	4.02	3.24	4.83
Tegaderm CHG cost	GBP	18.6	14.9	22.3

Running the model: base case

The following code runs a single model scenario, using the `evaluate` method of a decision node to evaluate each pathway from the decision node, shown in the table. This model did not consider utility, and the columns associated with utility are removed.

```
RES <- DT$evaluate()
```

Run	d1	Cost
1	Standard	176.7
1	Tegaderm	93.33

Running the model: probabilistic sensitivity analysis

Probabilistic sensitivity analysis is supported through the use of sampling model variables. The same call, with extra parameters is used to run the PSA:

```
PSA <- DT$evaluate(expected=F, N=100)
```

The first few runs of PSA are as follows (after reshaping the table to give one row per simulation, rather than one row per run, per strategy):

Run	Cost.Tegaderm	Cost.Standard	Difference
1	103.5	227.7	-124.2
2	128.2	192.3	-64.17
3	153.8	256.9	-103.1
4	106.2	204.2	-97.97
5	46.68	131.5	-84.82
6	137.8	191.1	-53.33

Run	Cost.Tegaderm	Cost.Standard	Difference
7	134.2	166.5	-32.3
8	107.4	180.2	-72.79
9	70.23	138.6	-68.4
10	119.8	213.6	-93.78

From PSA (100 runs), the mean cost of treatment with Tegaderm was 102.33, the mean cost of treatment with standard dressings was 176.27 and the mean cost saving was -73.94. The 95% confidence interval for cost saving was -127.98 to -15.38; the standard deviation of the cost saving was 29.48. Overall, 99% of runs found that Tegaderm was cost saving. These results replicate those reported by the manufacturer (saving of 77.76, 98.5% cases cost saving; mean cost of standard dressing 176.89, mean cost of Tegaderm 99.63).

References

- 1 Jenks M, Craig JA, Green W, *et al.* Tegaderm CHG IV securement dressing for central venous and arterial catheter insertion sites: A NICE Medical Technology Guidance. *Applied Health Economics and Health Policy* 2016;**14**:135–49.
- 2 Briggs A, Claxton K, Sculpher M. *Decision modelling for health economic evaluation*. Oxford, UK: : Oxford University Press 2006.