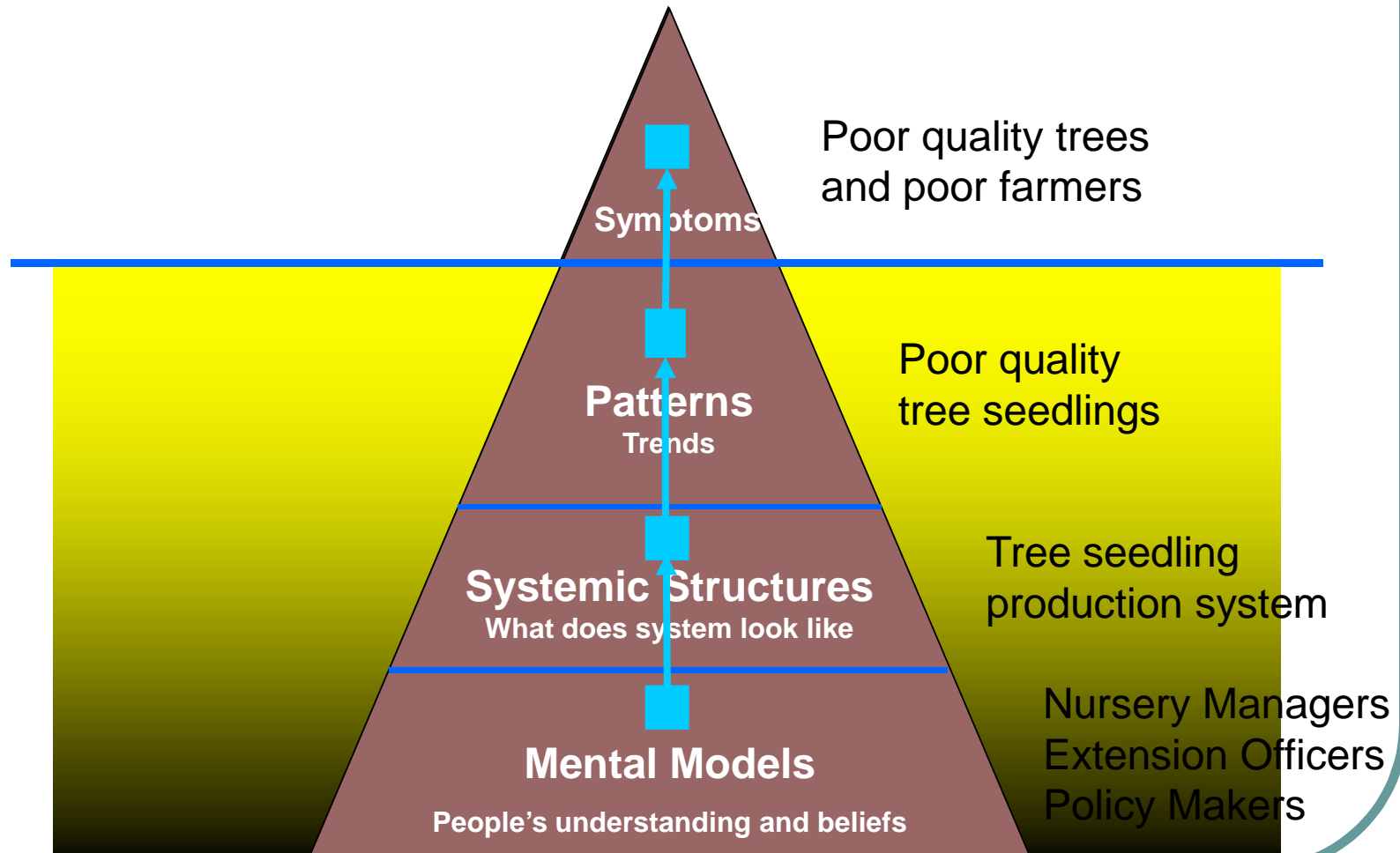


Applications of Participatory Systems Analysis and Bayesian Network models

Systems Thinking



Participatory Systems Analysis

- Systems Thinking with Stakeholders

Why?

- Reveal peoples mental models, i.e. how they think
- Expose and share ideas and views – collaborative learning
- Avoid parachuting in solutions and being told to 'Get-Lost' – empower stakeholders, improve adoption

Bayesian Networks

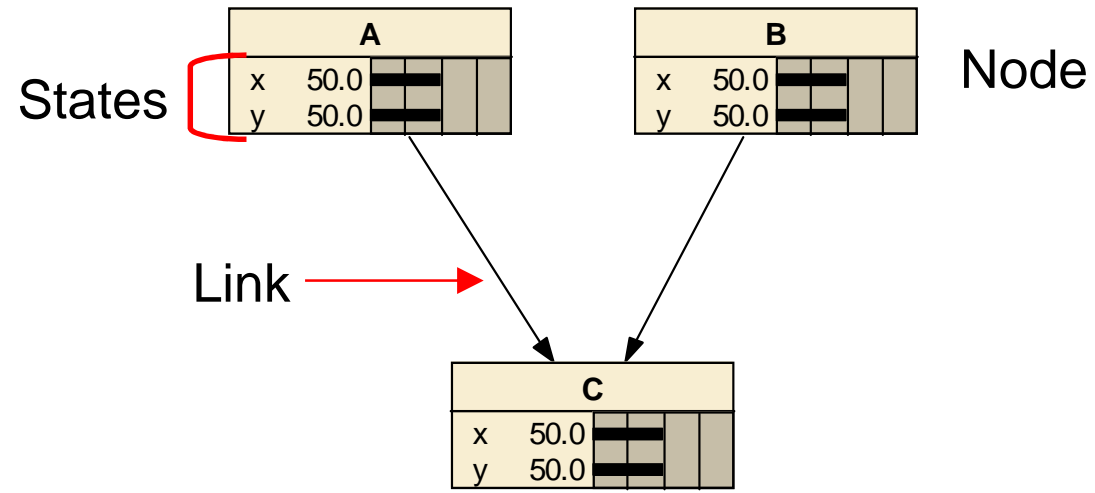
- Quite a handy tool for PSA

Why?

- Visual – easy of non-modellers
- Flexible – can link both qualitative and quantitative factors
- Uncertainty – probabilities
- Scenario, Diagnostic and Sensitivity Analysis – performed rapidly

What is a Bayesian Network?

- Two components
 - Structure (Directed Acyclic Graph): Nodes, States, Links
 - Probabilities
- Links between variables represent causal relationships (as with a conceptual model)
- Probabilities are used to quantify the relationship between variables
- Bayes Theorem



Node: **C**

Chance ▾ % Probability ▾

A	B	x	y
x	x	80.000	20.000
x	y	50.000	50.000
y	x	50.000	50.000
y	y	20.000	80.000

Conditional Probability Table (CPT)

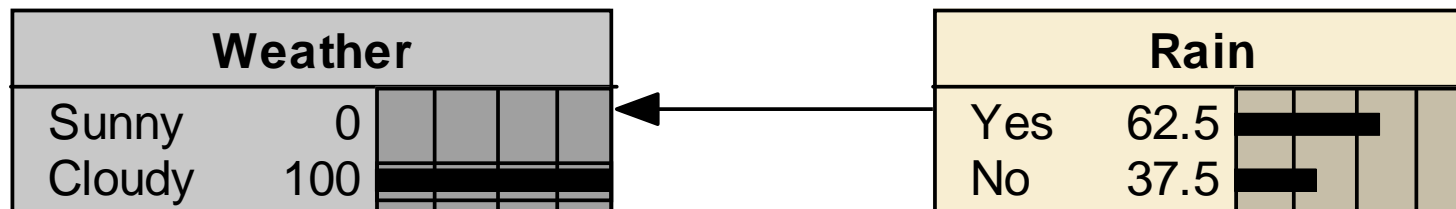
Bayes Theorem

$$P(B|A) \times P(A) = P(A|B) \times P(B)$$

$$P(B|A) = \frac{P(A|B) \times P(B)}{P(A)}$$

$$P(\text{Rain}|\text{Cloudy}) = \frac{P(\text{Cloudy}|\text{Rain}) \times P(\text{Rain})}{P(\text{Cloudy})}$$

$$P(\text{Rain}|\text{Cloudy}) = \frac{100 \times 50}{80} = 5000/80 = 62.5 \%$$



Rain	Sunny	Cloudy
Yes	0.000	100.00
No	40.000	60.000

Yes	No
50.000	50.000

Bayes



Rev. Thomas Bayes
(1702-1761)

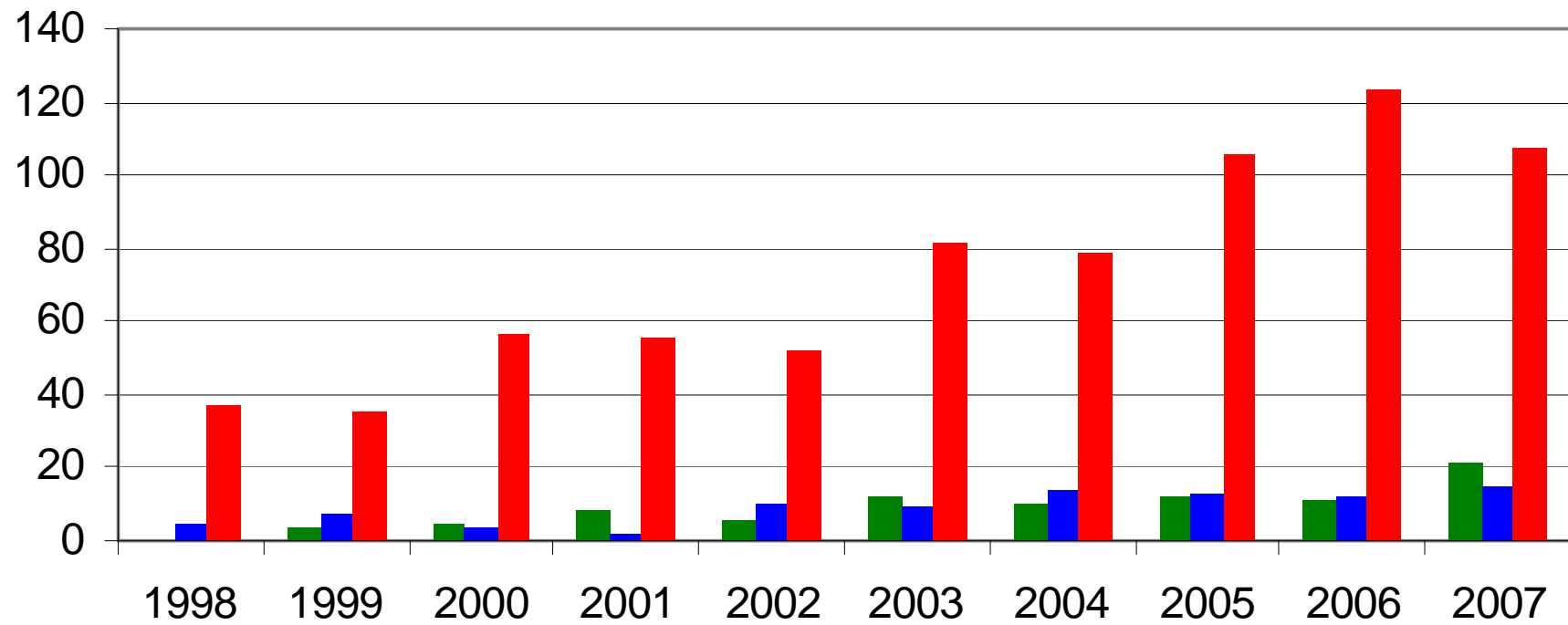
Then



Now

Modelling Issues	Process/ System Dynamic Models	Bayesian Networks	Agent-based Models	Mathematical/ Statistical Models	Expert Systems
Can model cyclical processes and feedback loops?	X		X		
Can model non-linear system dynamics?	X	X	X	X	
Can easily accommodate and represent uncertainty in predictions?		X			X
Can be used to model emergent behaviour and self-organisation?			X		
Can model system dynamics continuously over time?	X		X	X	
Can model interacting processes operating at different temporal and spatial scales?	X		X		
Can be used for prediction and scenario analysis?	X	X		X	X
Can be used for back-casting and diagnostic analysis?		X			X
Can be used in data poor situations?		X	X		X
Are simple for managers and non-modellers to use and interpret?		X			X
Have relatively low development cost?		X			X
Are relatively easy to update?		X			X

Number of BN articles in Current Contents - different subject areas



Red = Engineering and Computer Science
Blue = Clinical Medicine
Green = Ag, Biology and Env. Science

ABNMS2010 Conference Talks

Bayesian network modelling from the **stakeholders perspective**

Bayesian Networks as a **Complex System tool** in the context of a major industry and University project

Predicting a 'tree change' in Australia's tropical savannas: Combining different types of models to understand **complex ecosystem behaviour**

Combining state and transition models with **dynamic Bayesian networks**

BNs for **Anomalous Vessel Detection**

Bayesian modelling for the **Assessment of Proposed Indirect Potable Reuse Schemes**

Is dingo management the key? Using a Bayesian Belief Network to **assess the impact of an apex predator on biodiversity conservation** in the Australian arid zone

Modelling invasive risk of a changing landscape: Using BBNs to **target areas for effective weed management**

Bayes Nets as a method for **analysing return for investment in fire management planning**

Interpreting Risk and Motivational Context: **Importing New Zealand Apples into Australia**

A Bayesian Belief Network for **Legal Decision Support** in Owners Corporation Cases

Consumer Credit Scoring: Building a better predictor using causal Bayesian networks

Object Oriented Bayesian Networks: Designing for simplicity and integration

Build collaborative models or capacity? Reflections from two years on

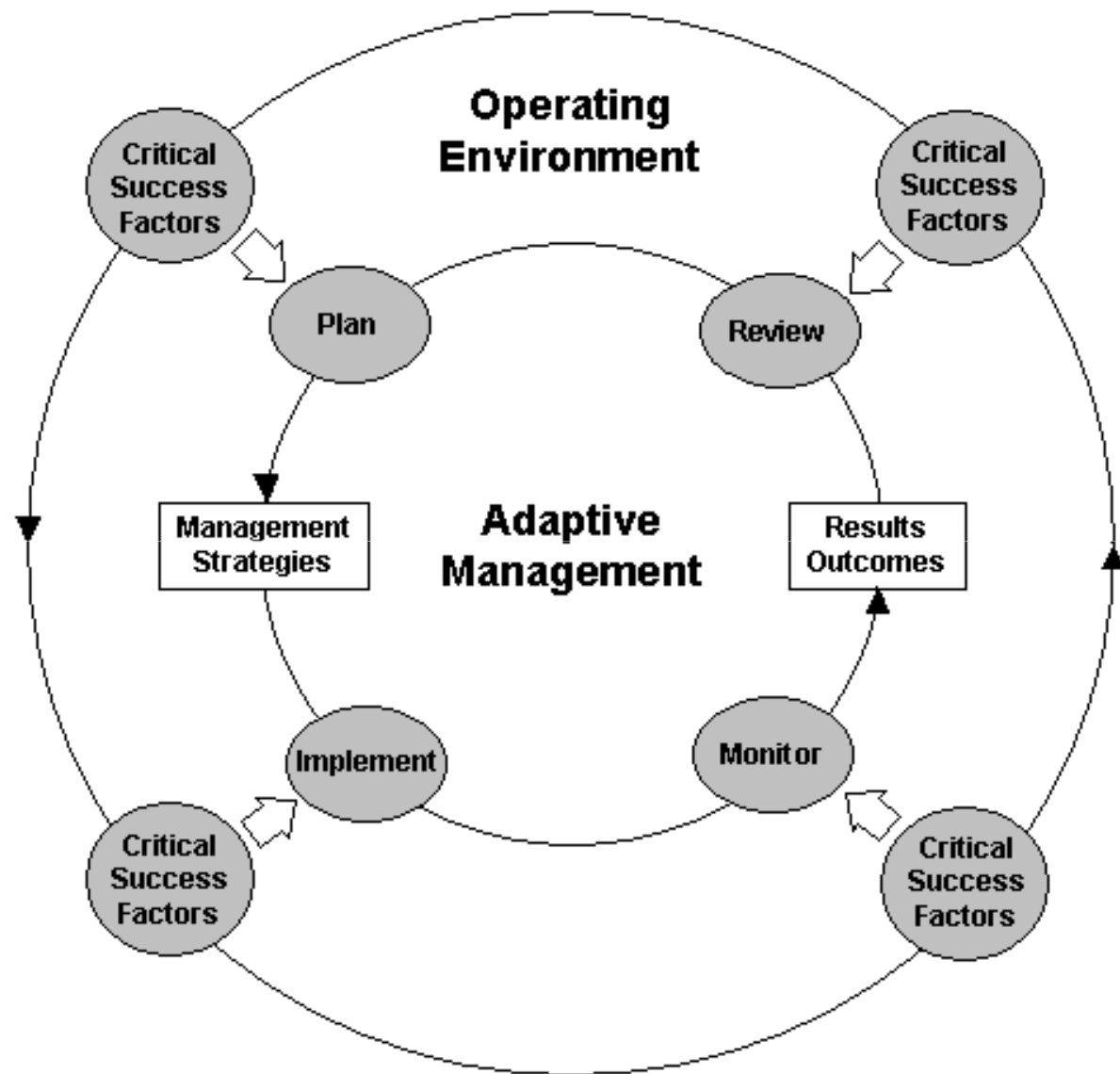
Example 1 - QPWS

The Problem:

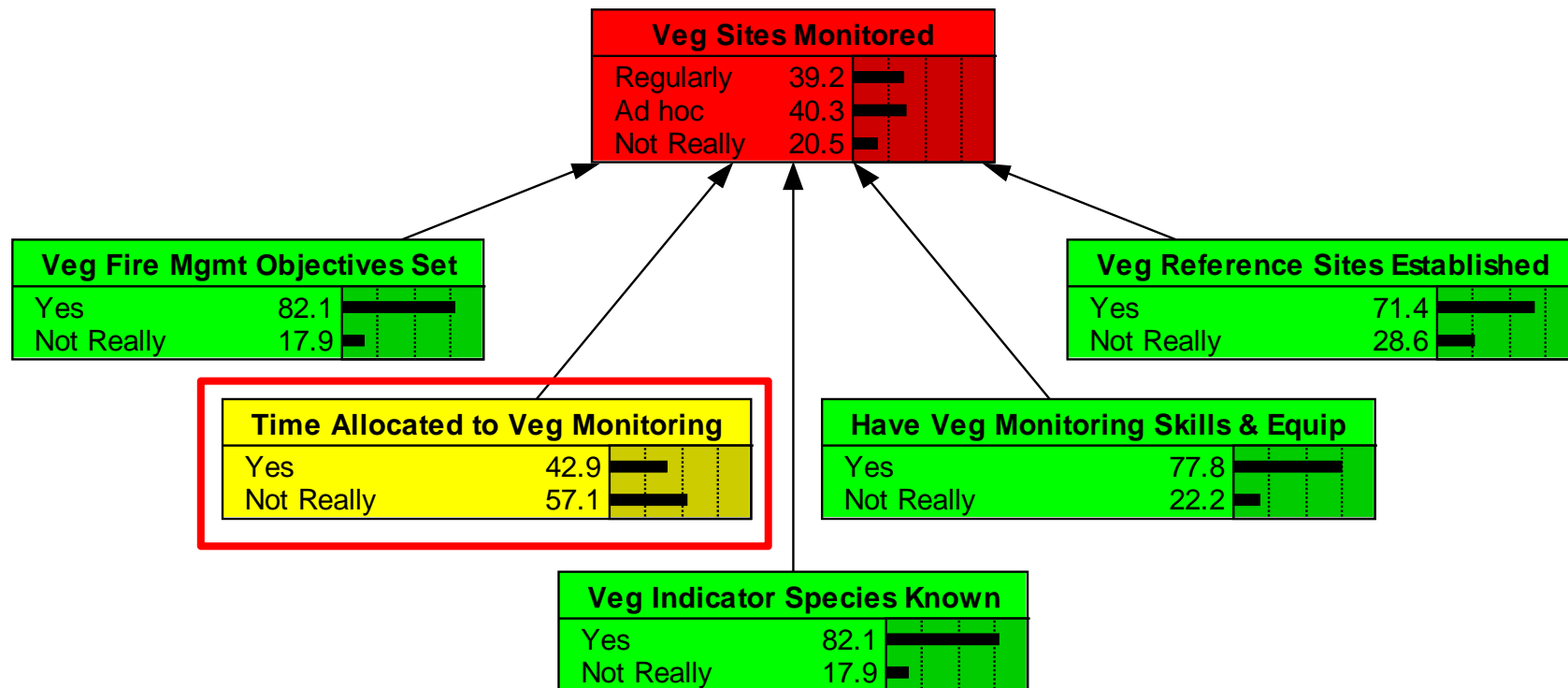
- Had a policy – Adaptive Management of Fire on conservation reserves
- Spent heaps and money on implementing this policy
- But Adaptive Management still wasn't happening

What we did:

- Systems analysis with park rangers

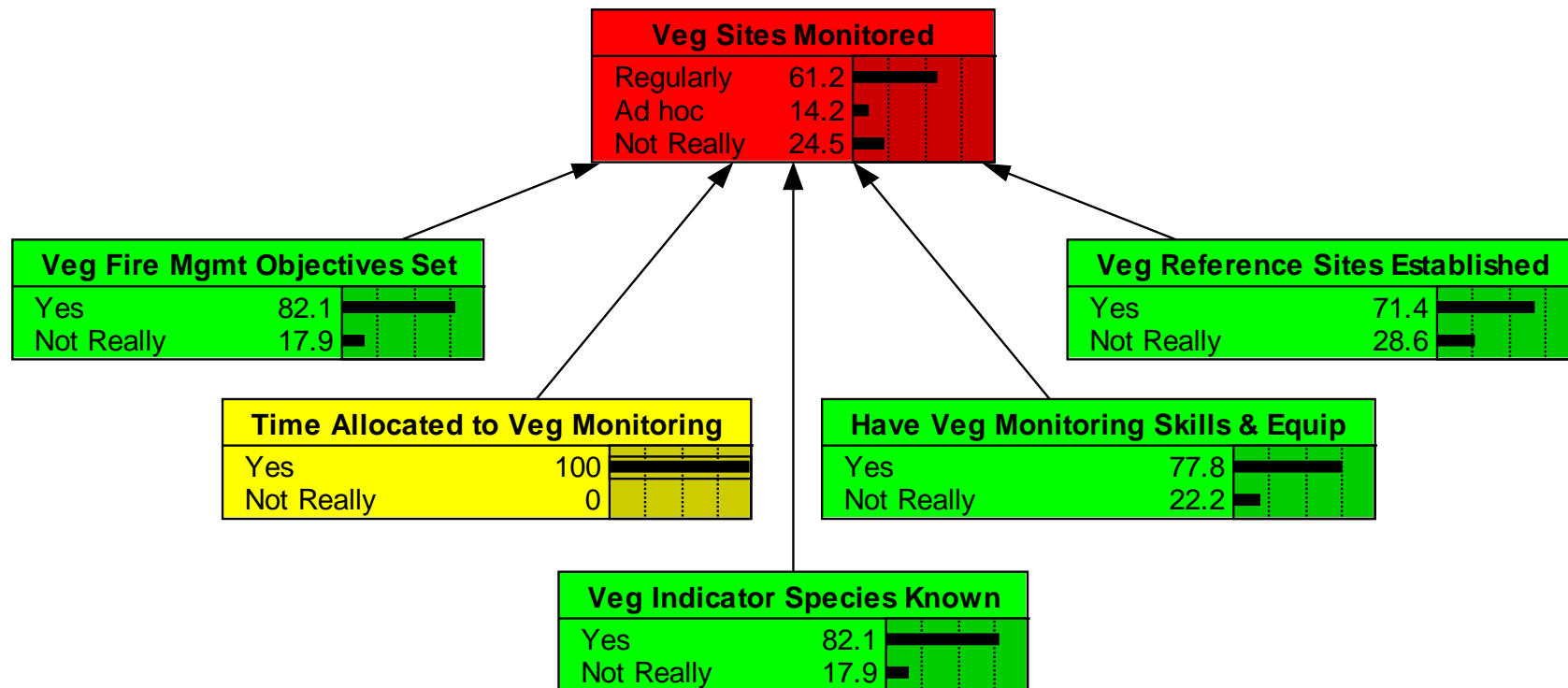


Performance



Too busy cleaning dunnies

Intervention Assessment



Example 2 – Tree Seedlings Philippines

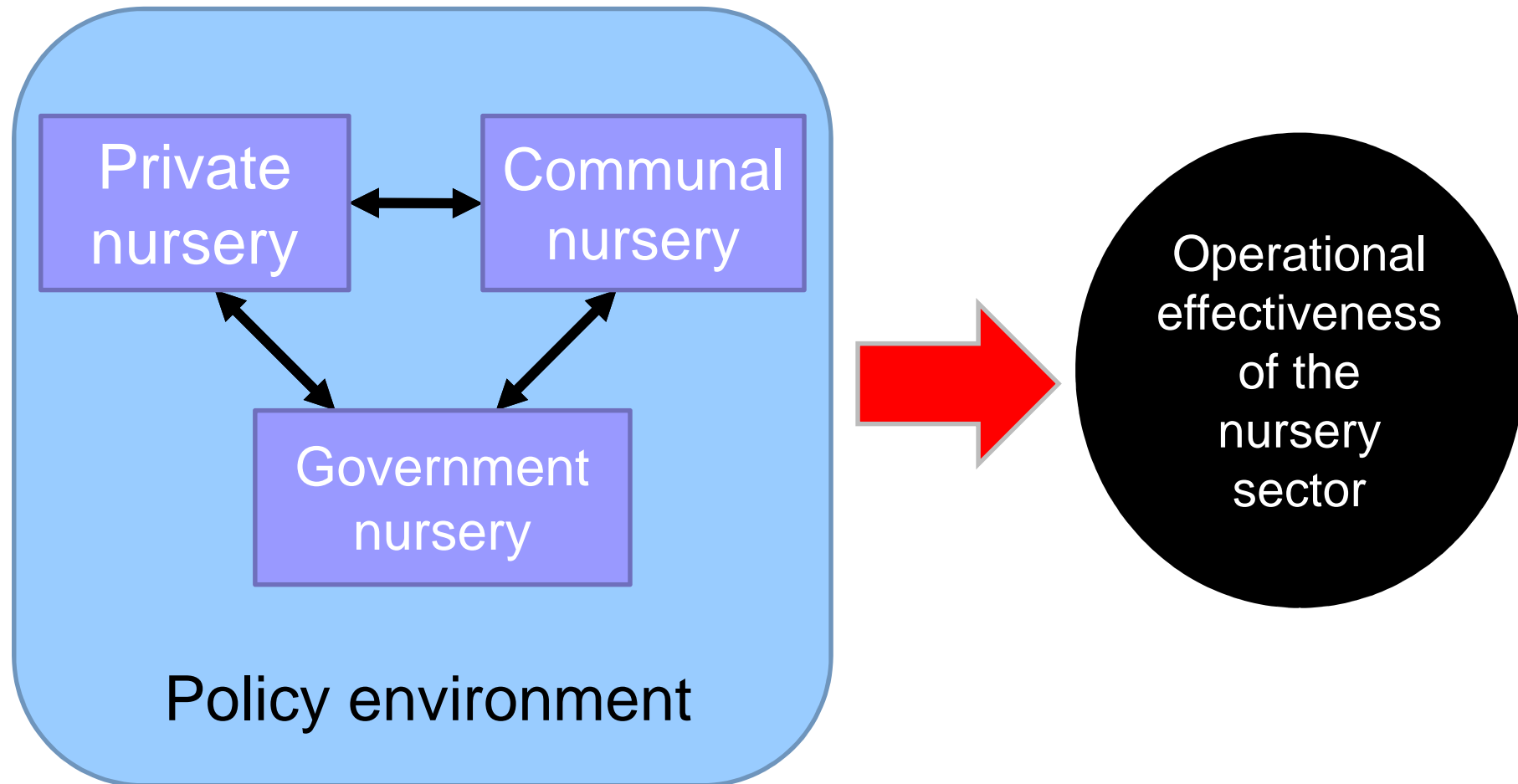
The Problem:

- No quality protocols in most nurseries
- Poor seedlings – poor trees
- Low financial viability – not able to sustain nursery operations

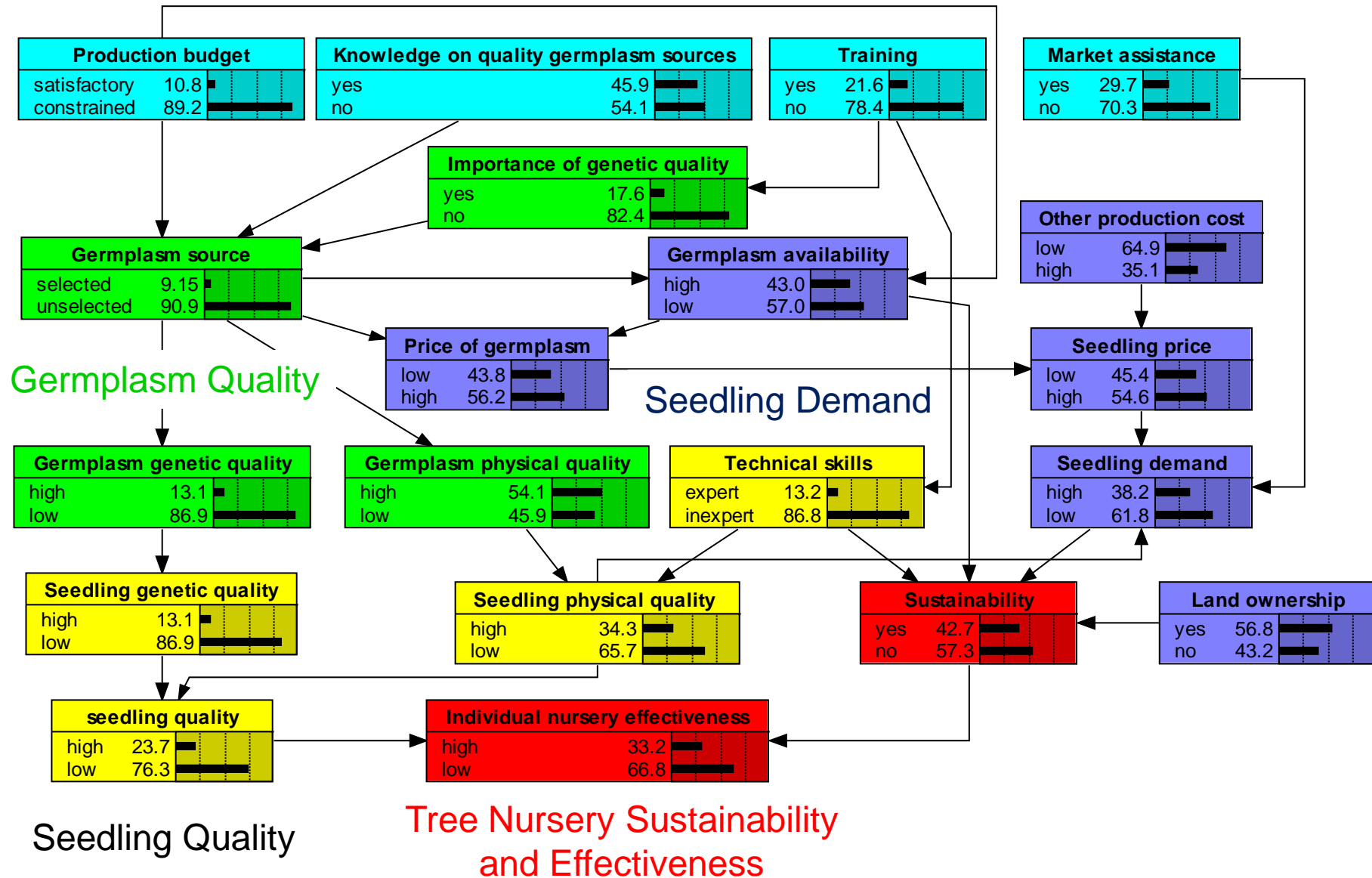
What we did:

- Systems analysis with nursery managers, extension officers and government

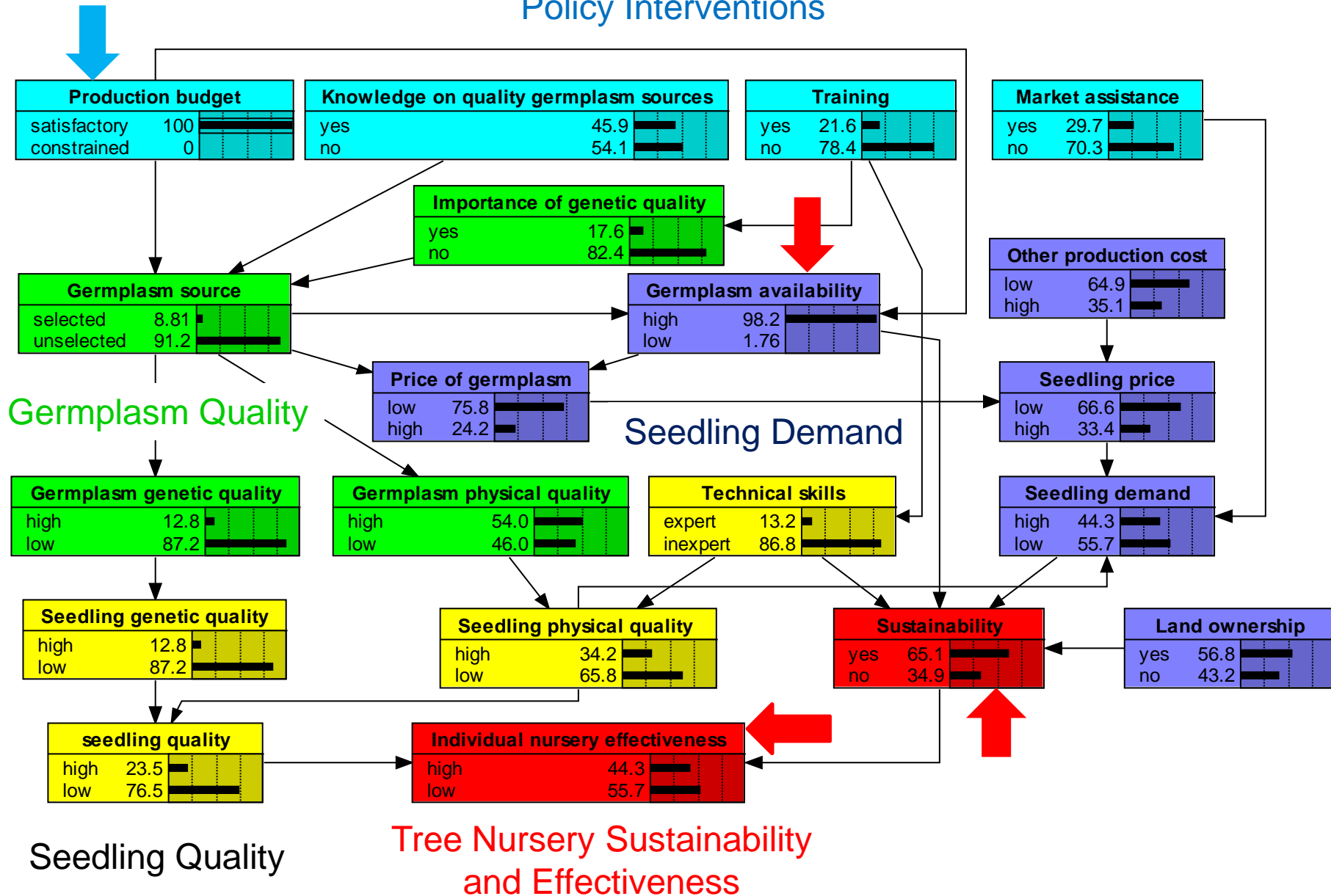
Operational effectiveness of the forestry nursery sector



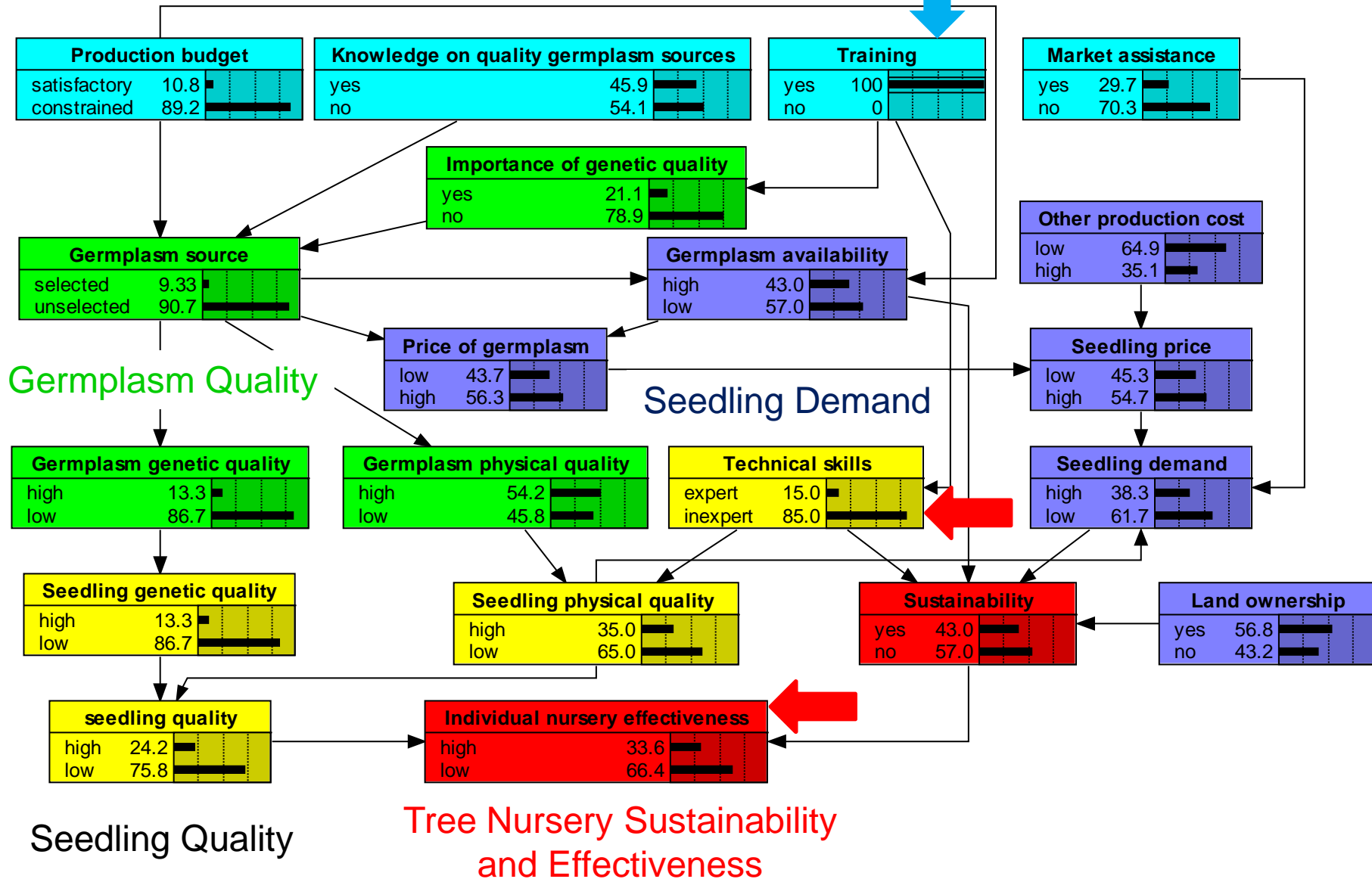
Policy Interventions



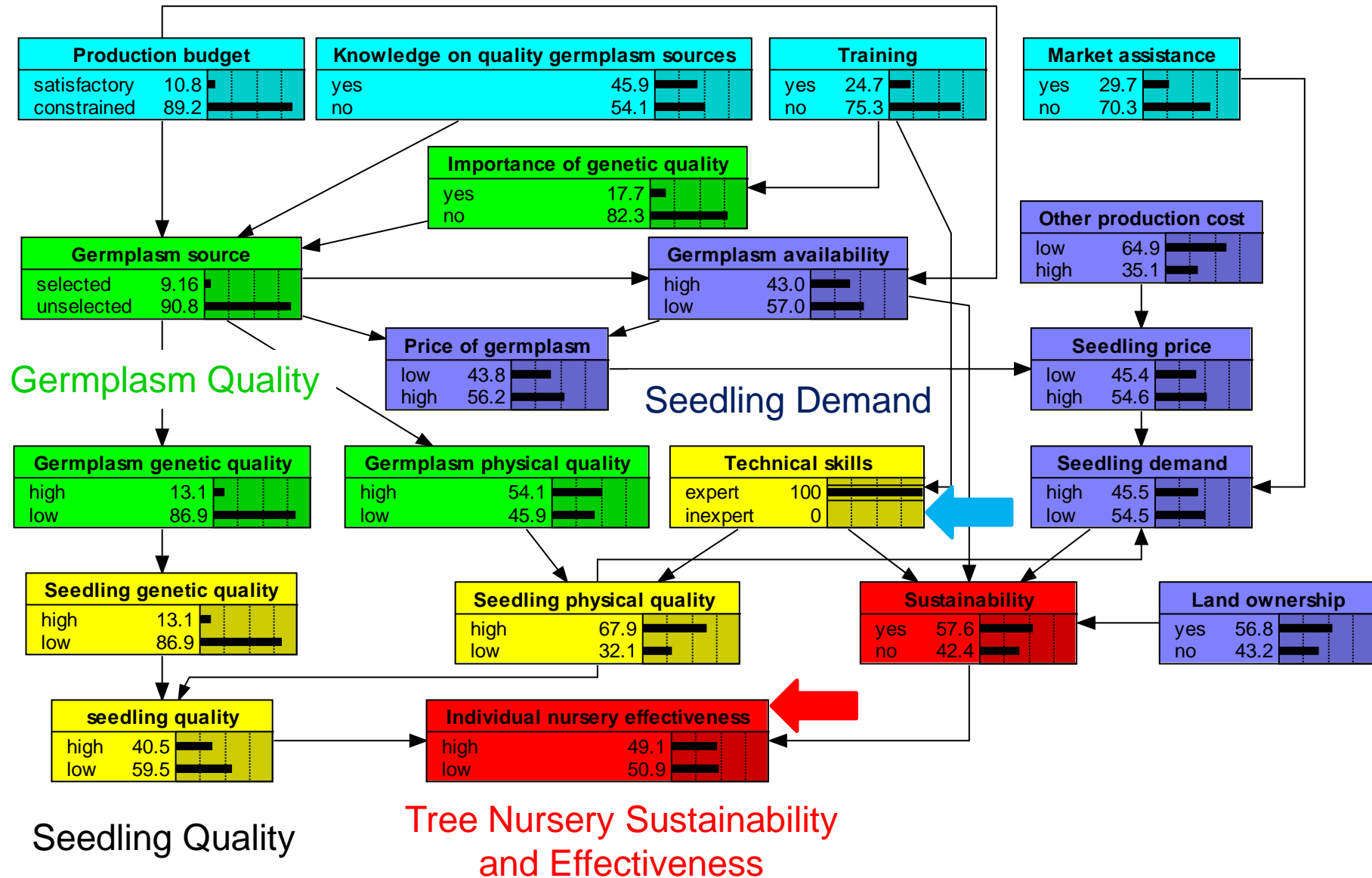
Policy Interventions



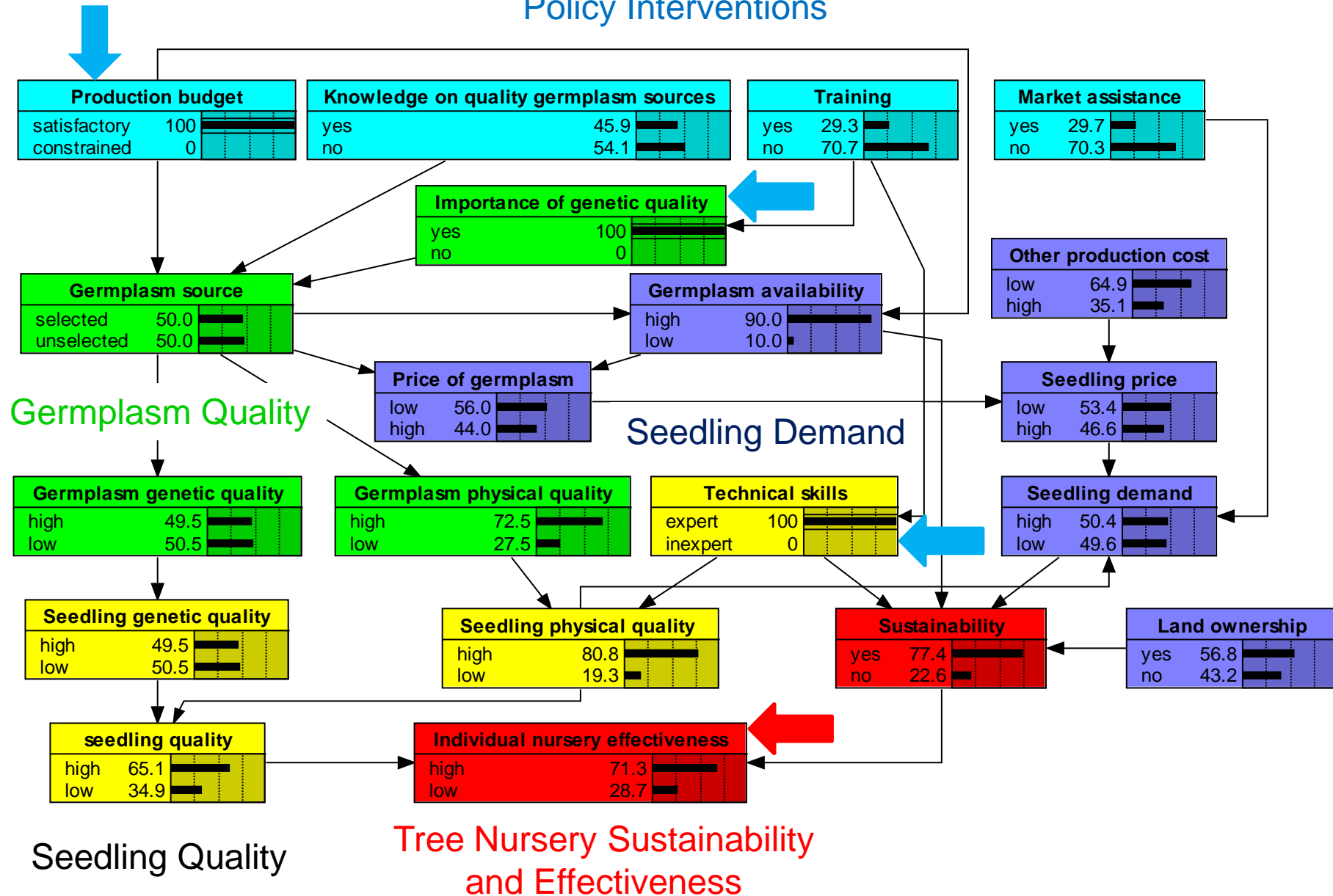
Policy Interventions

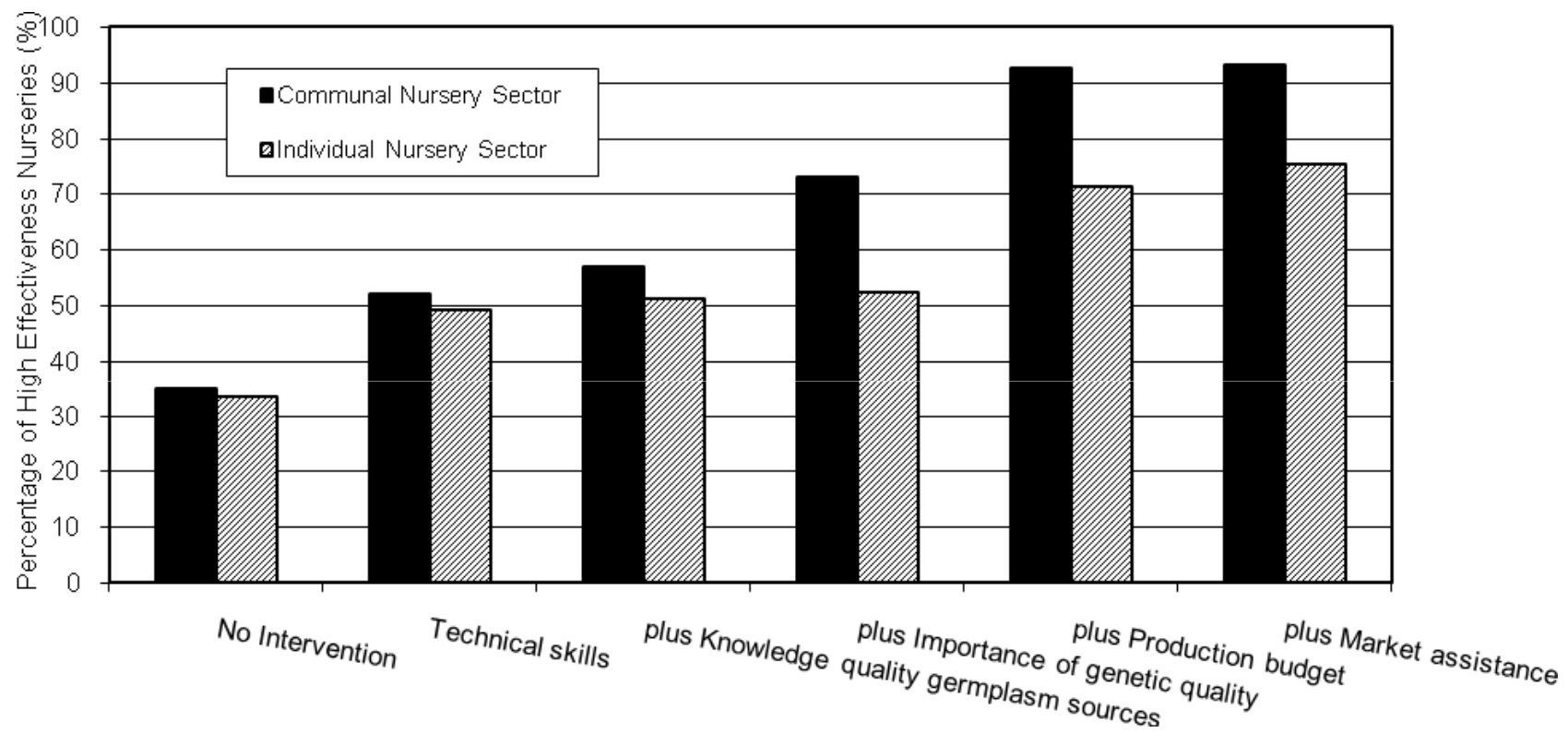


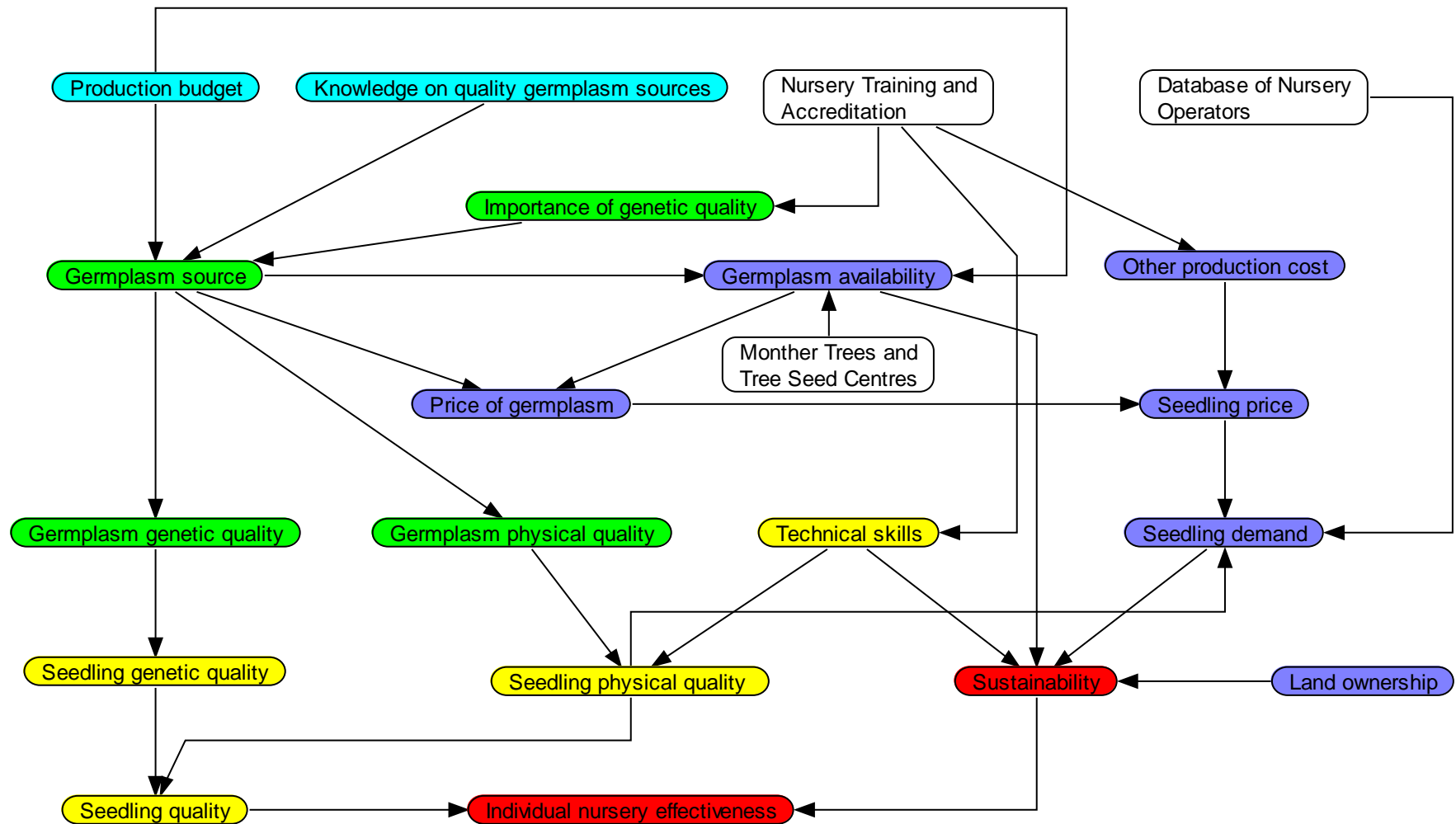
Policy Interventions



Policy Interventions







- Interventions

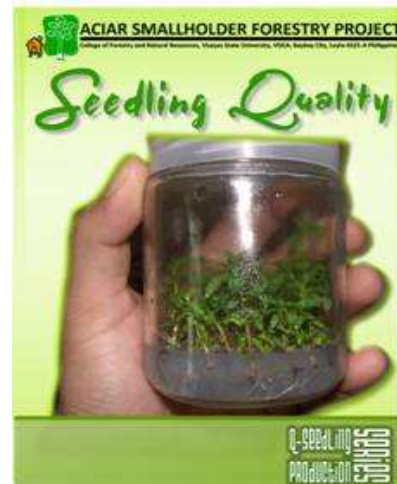
- Forestry nursery accreditation

Participatory research and partnership with LGUs

Development of nursery accreditation policy and guidelines



- Nursery accreditation (cont.)
Capacity building (series of training, nursery manuals and videos, establishment of demo nurseries)



- Improvement of germplasm availability

 - Establishment of tree seed centers

 - Database of germplasm sources

- Developed database of nursery operators

 - Name, contact number, address

 - Species produce and selling price

 - Marketing agreement



Example 3 – Parkinsonia in Queensland

The Problem:

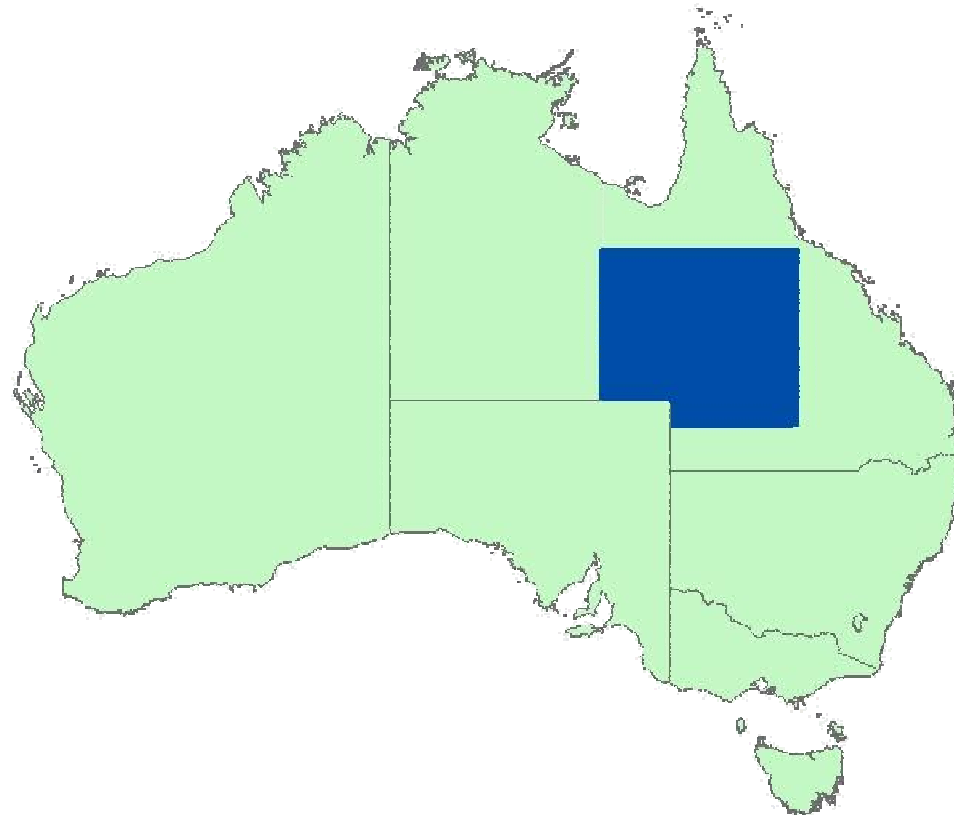
- Parkinsonia is an emerging weed – current models of potential distribution are based on climate matching
- Grows in rangeland and desert areas – coarse and limited data sets
- Need to make management decisions to contain spread

What we did:

- Systems analysis with weed ecologists, extension officers and graziers



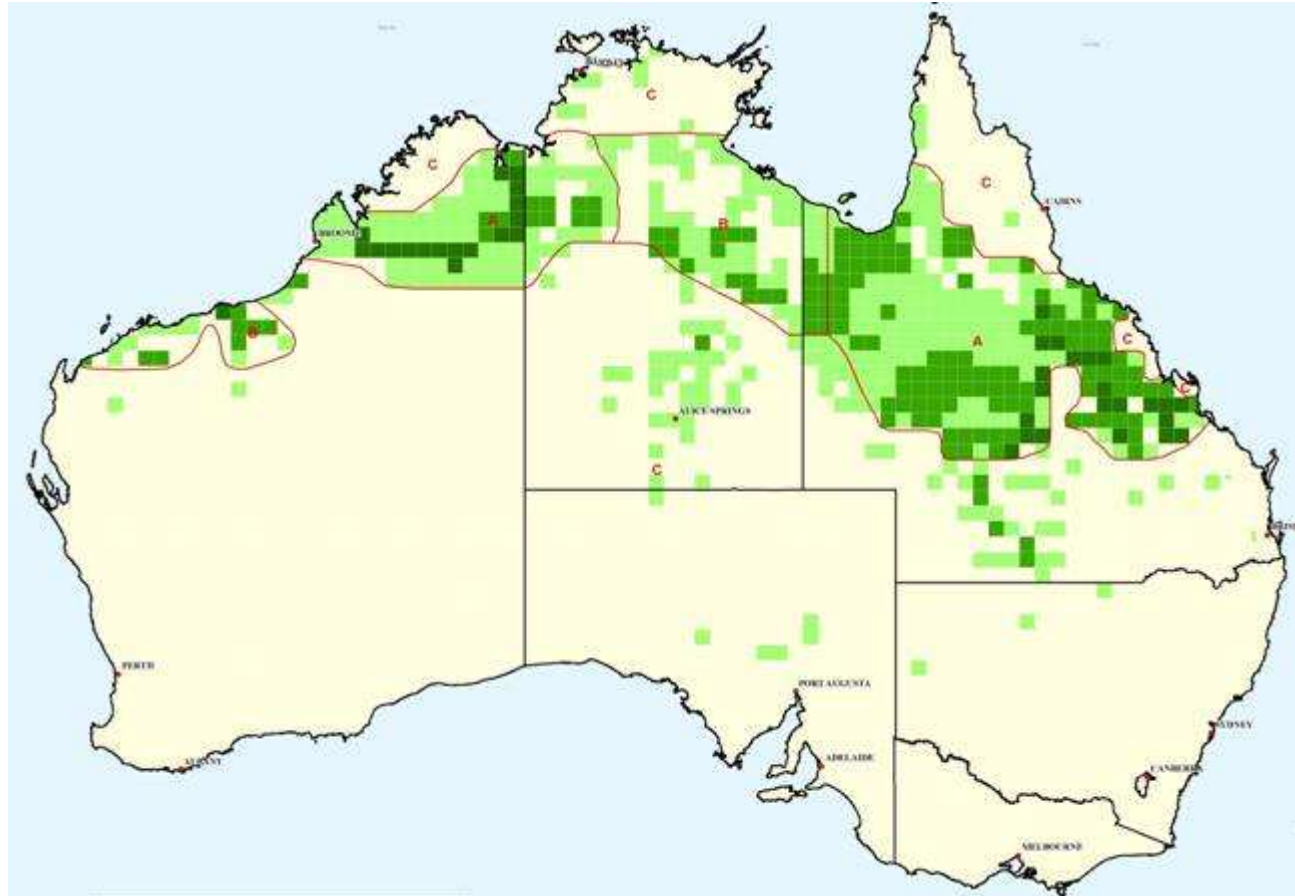
Desert Channels Region



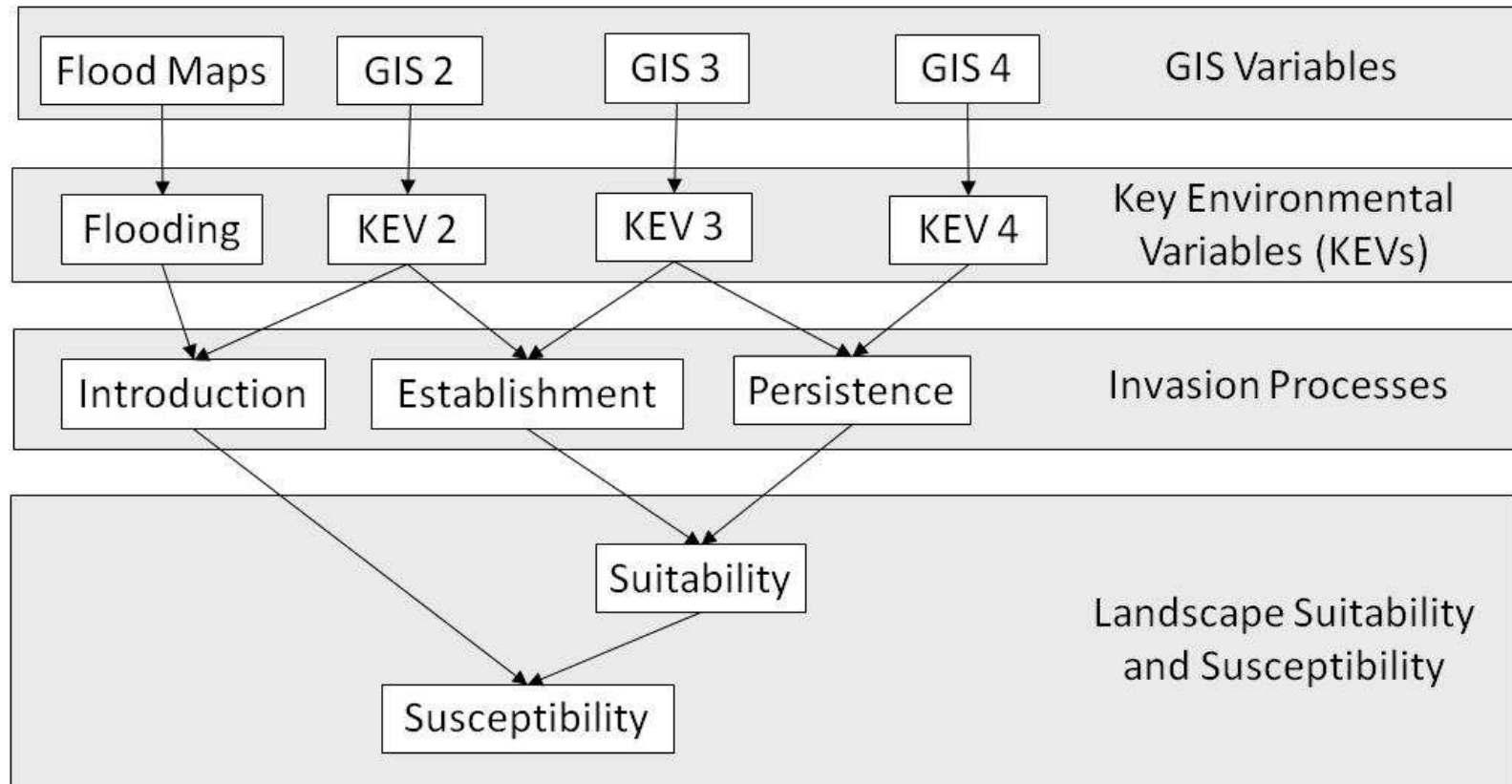
Area = 700 000 km² (Area of Austria = 84 000 km²)



Results from Climate Matching



Modelling Approach









REQUIREMENTS

VARIABLES

INTRODUCTION

2. VEHICLE MOVEMENT

1. FIXED EVENTS

3. VEHICLE MOVEMENT
2. HUMAN ERROR

Fixed - vehicle
not the human

Fixed - vehicle
not the human

Fixed - vehicle
not the human

Fixed - vehicle
not the human

Fixed - vehicle
not the human

Fixed - vehicle
not the human

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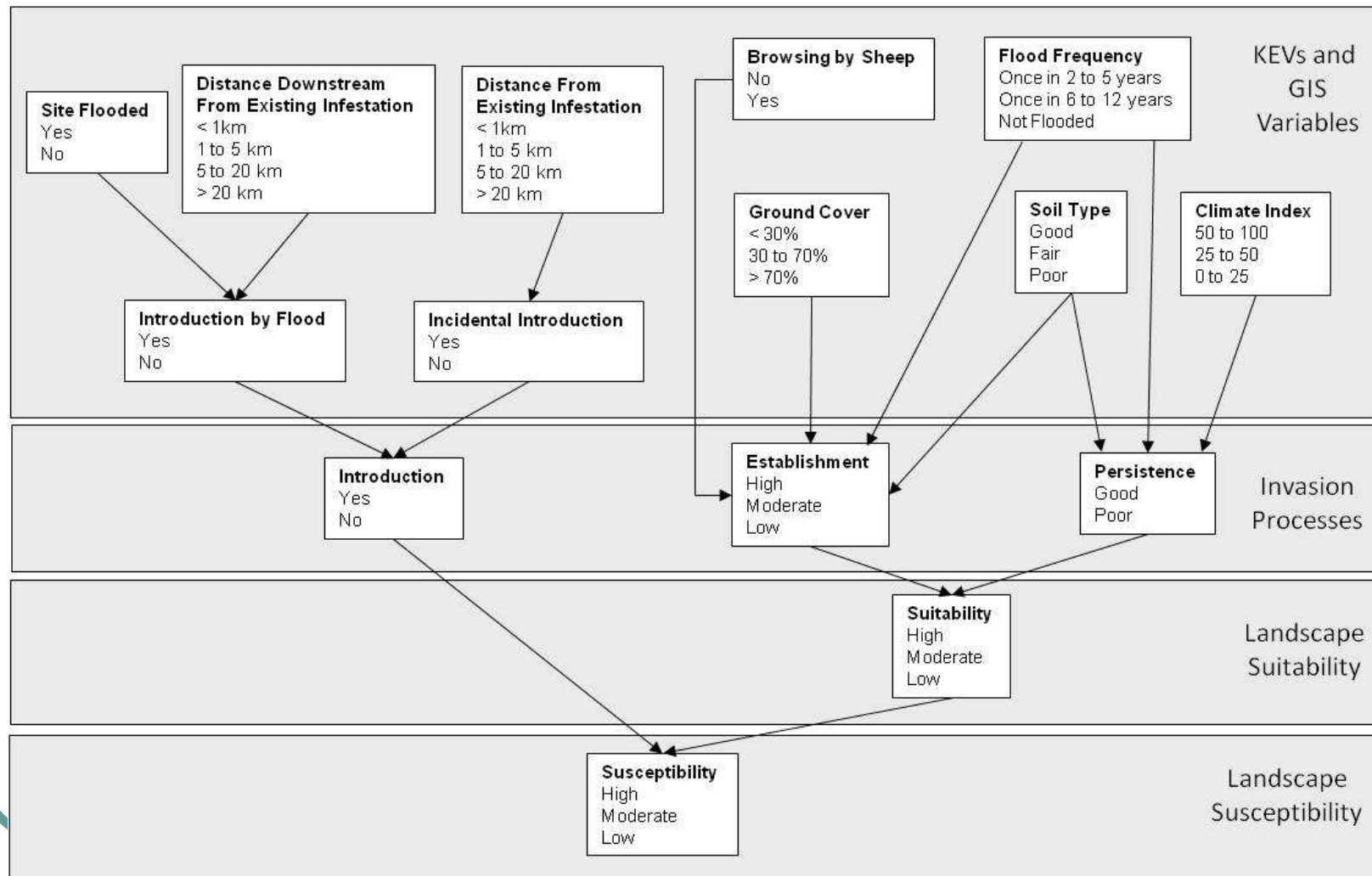
SUSCEPTIBILITY

ESTABLISHMENT

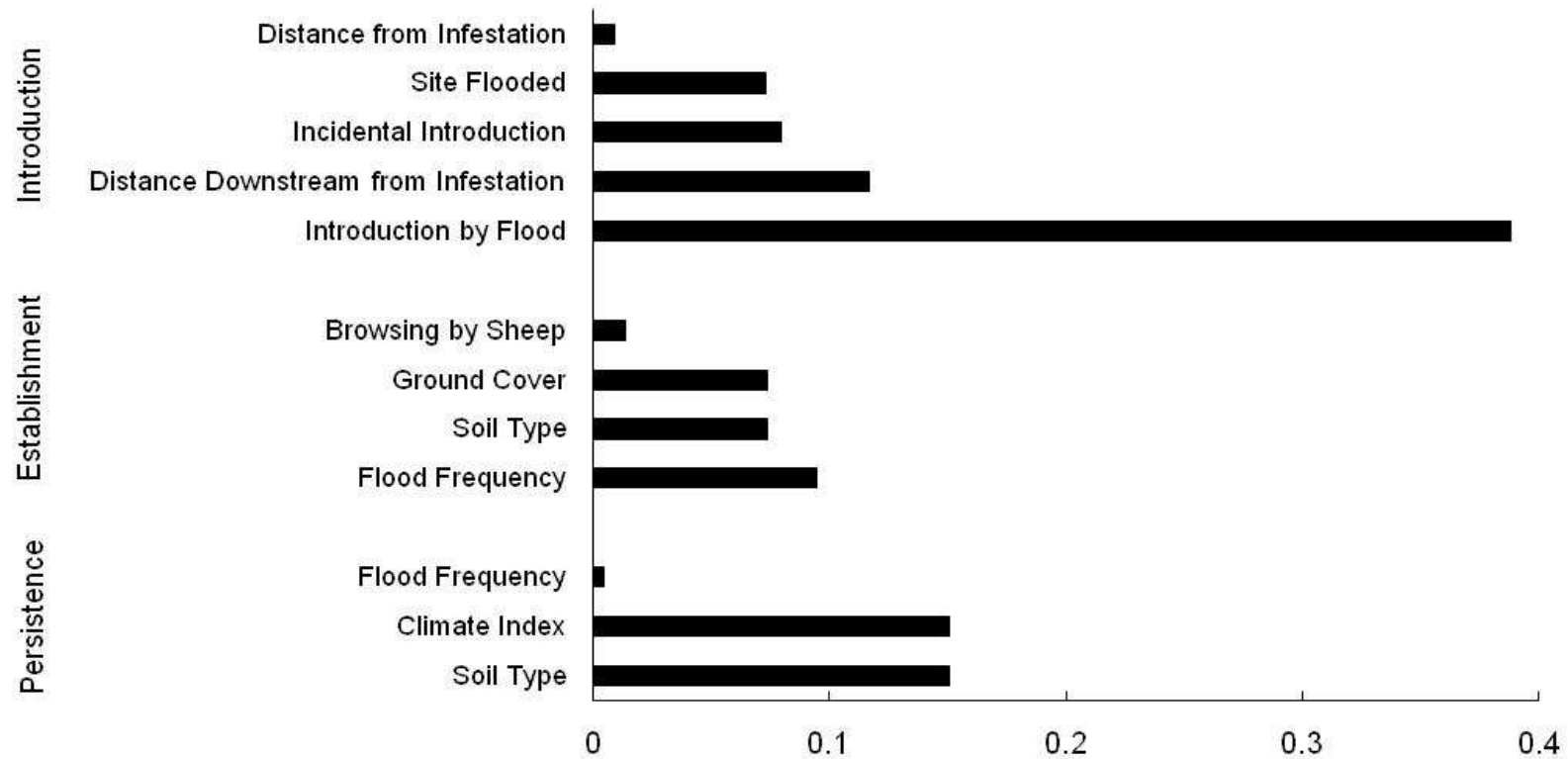
SURVIVAL



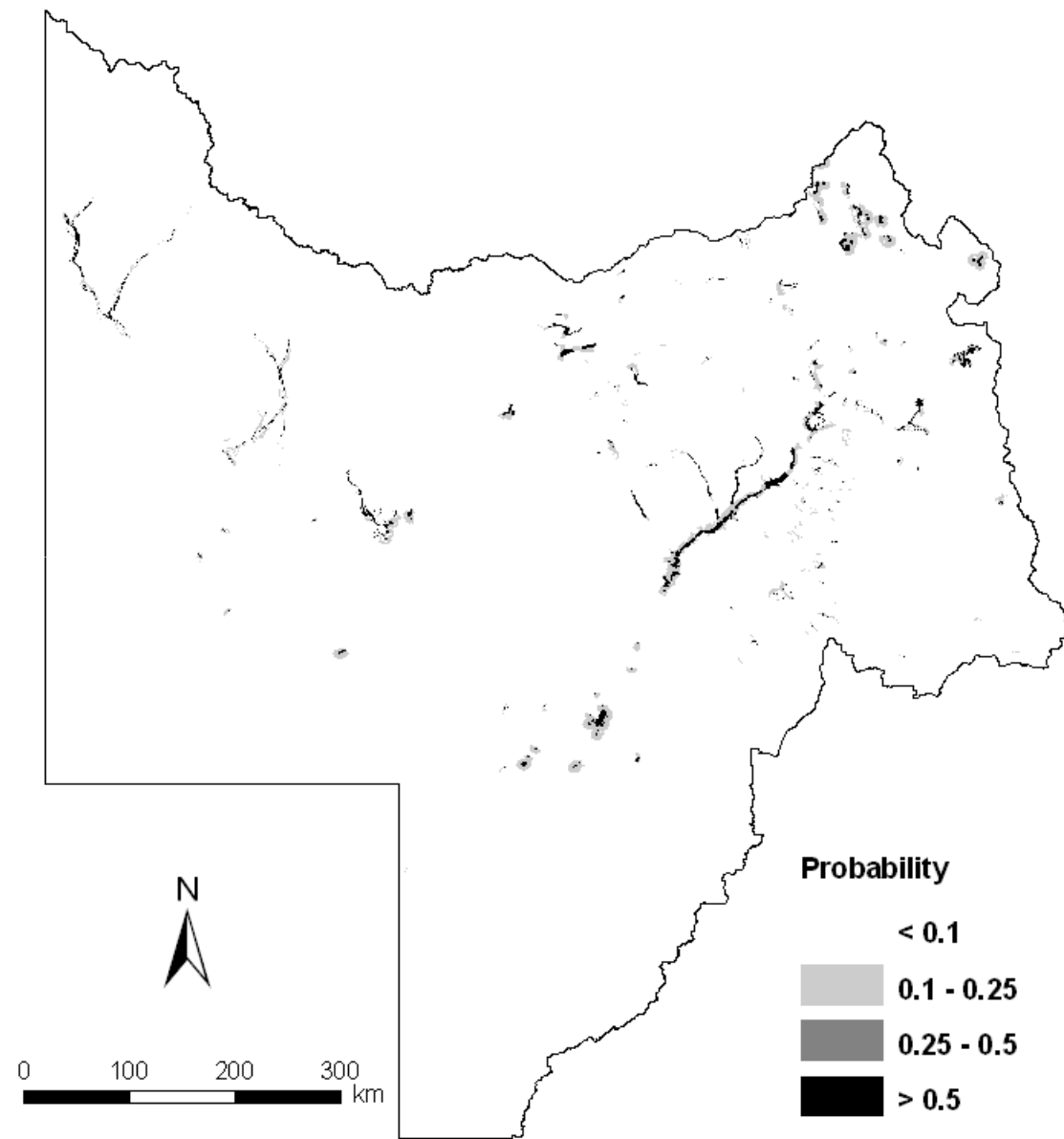
The Model



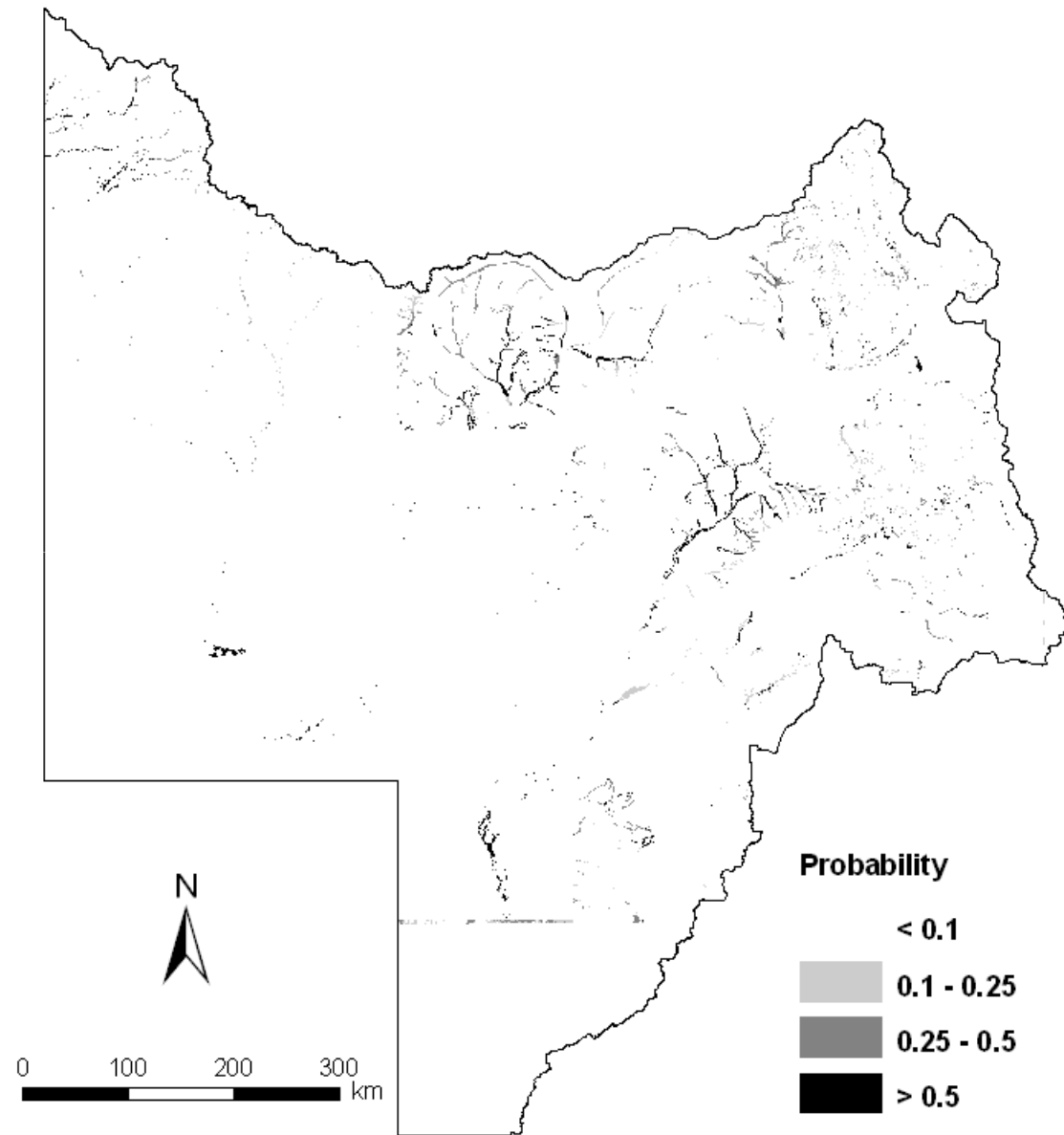
Sensitivity Analysis



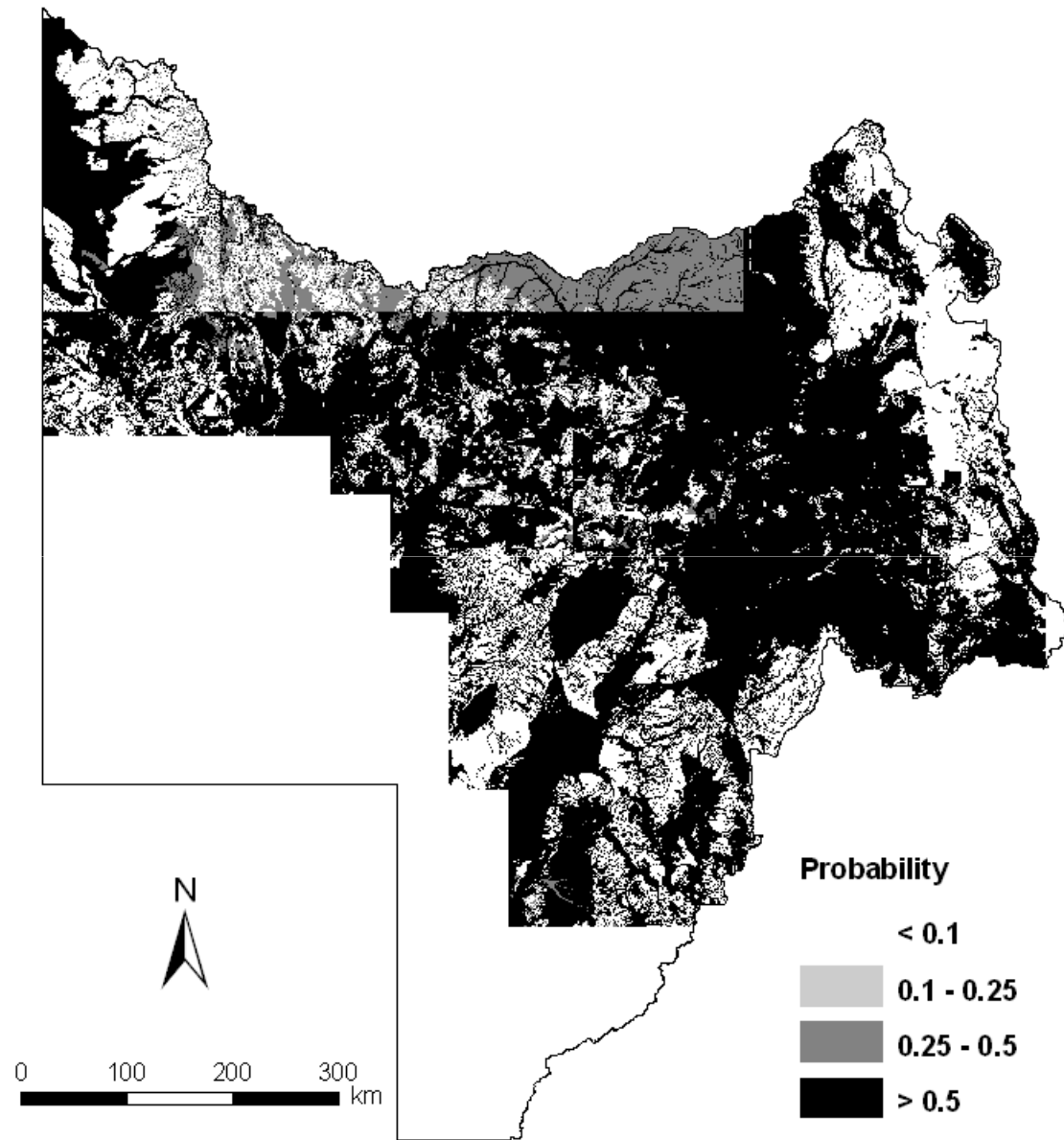
Introduction



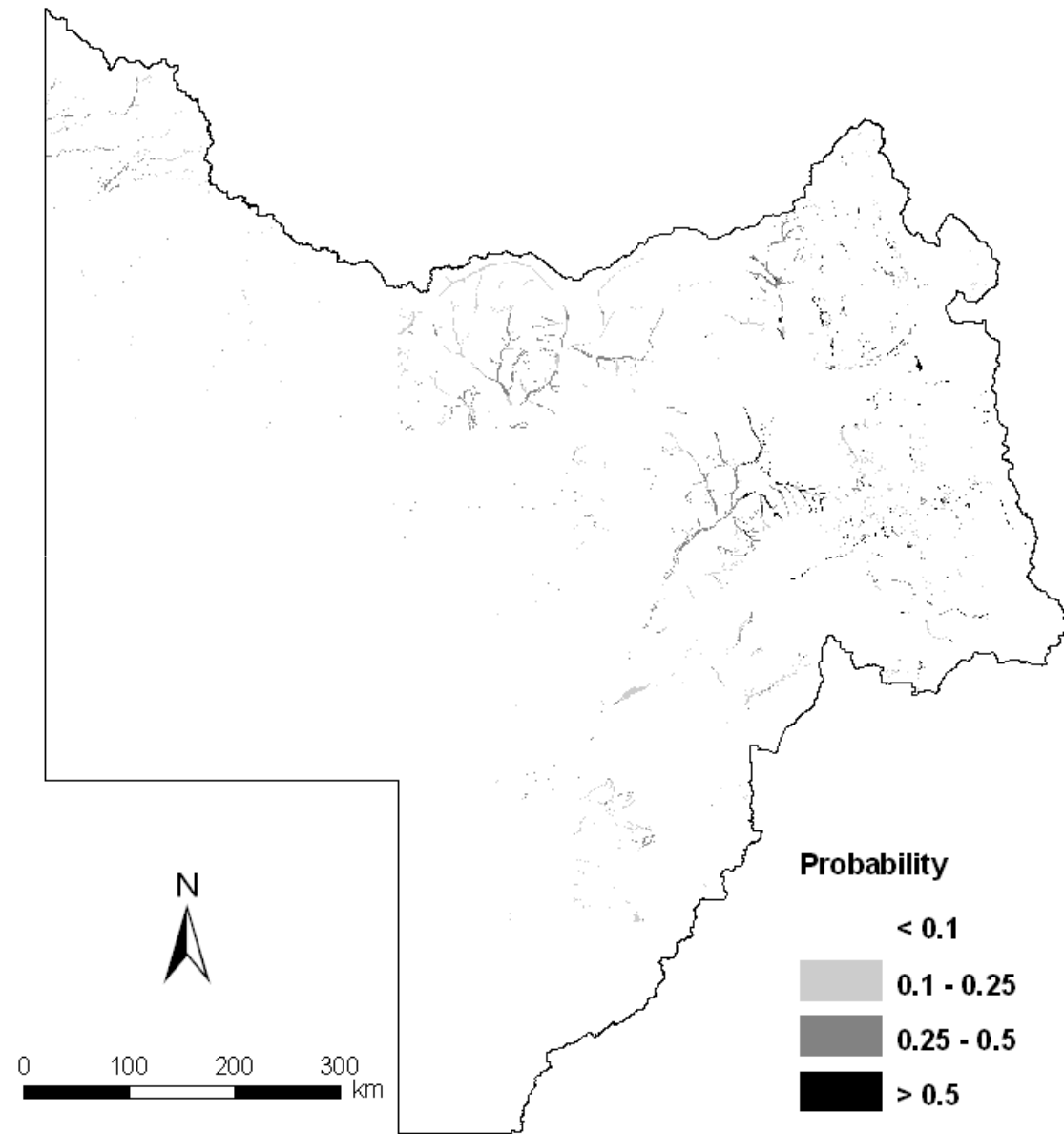
Establishment = High



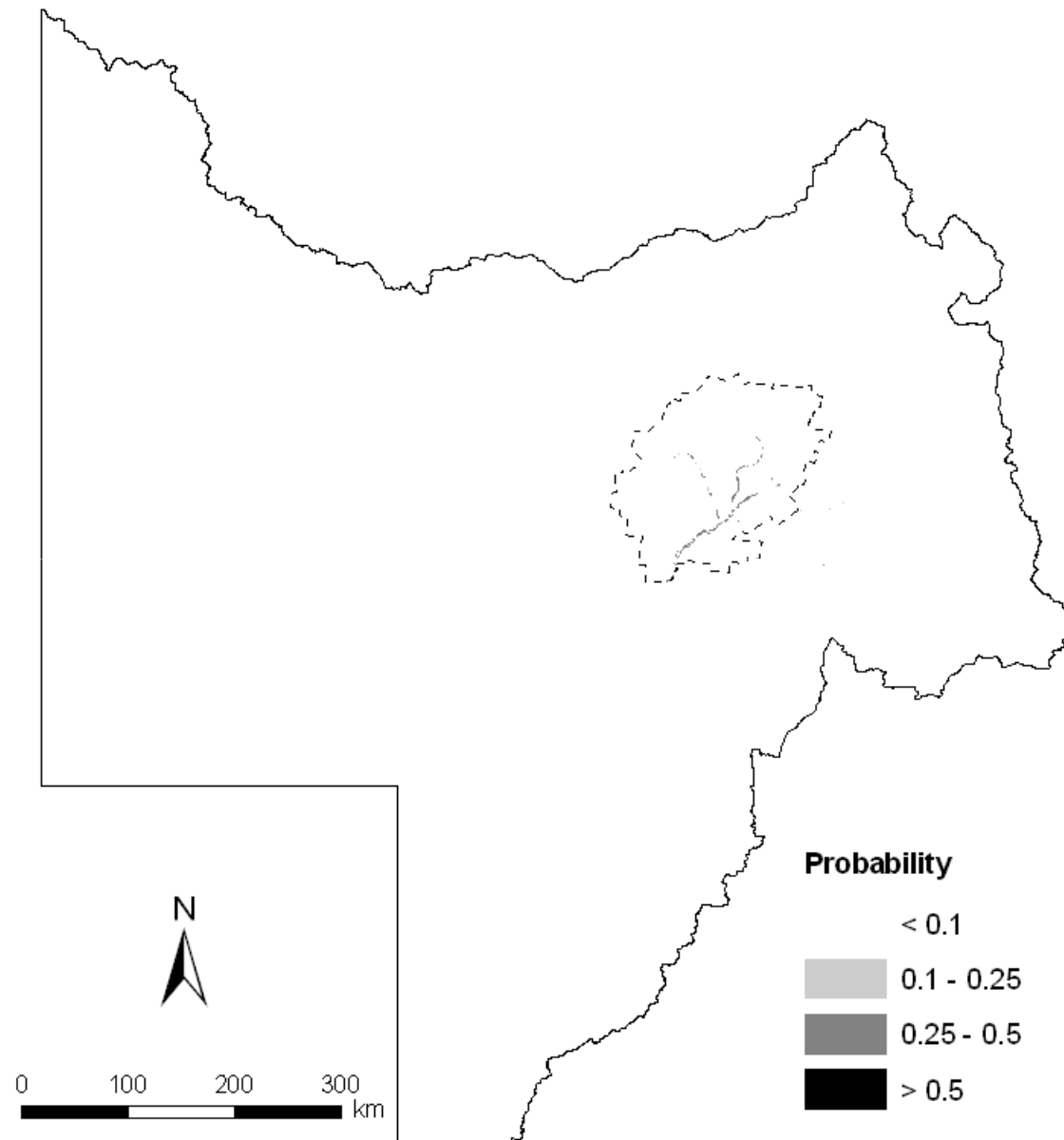
Persistence = Good



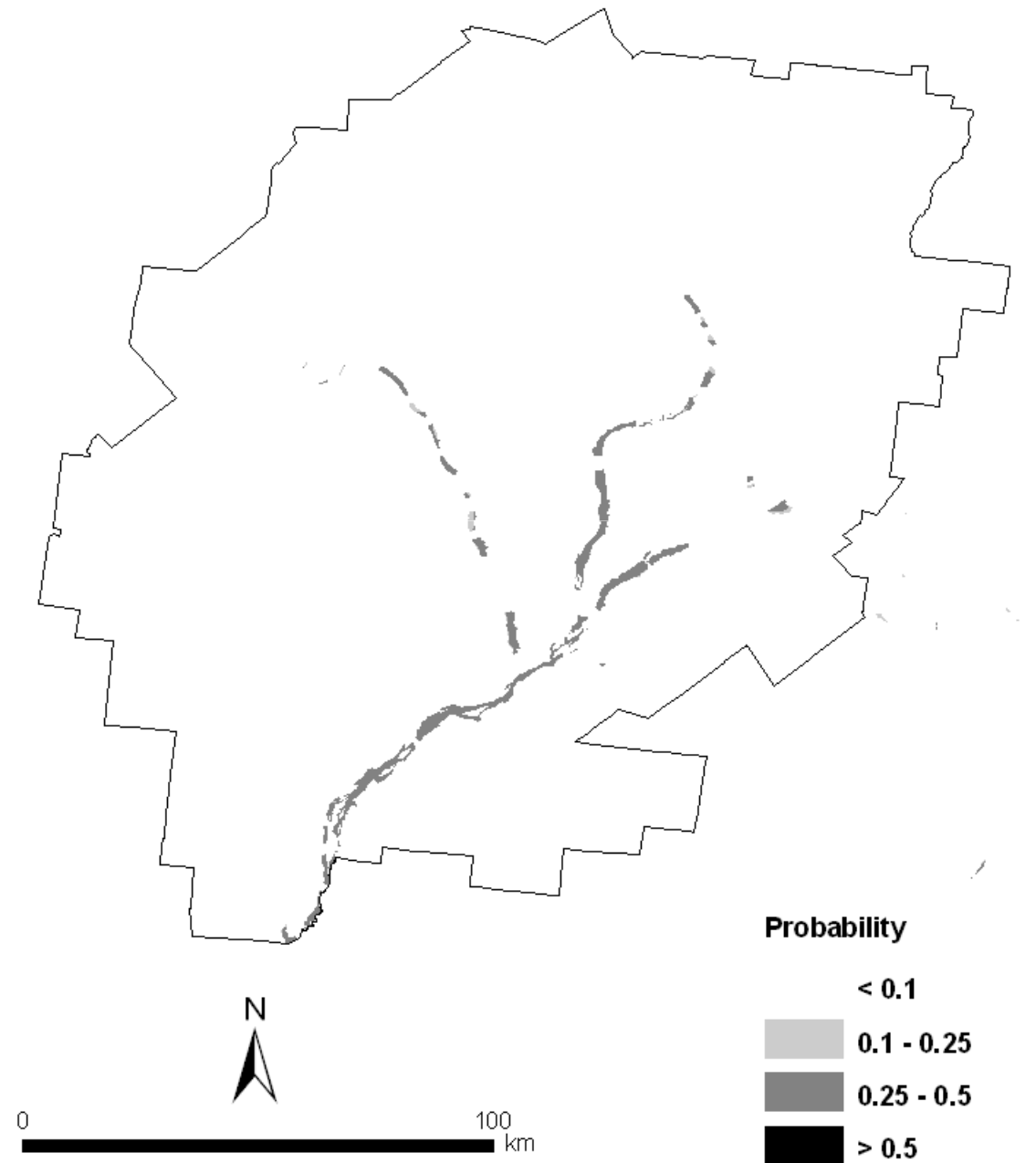
Suitability = High

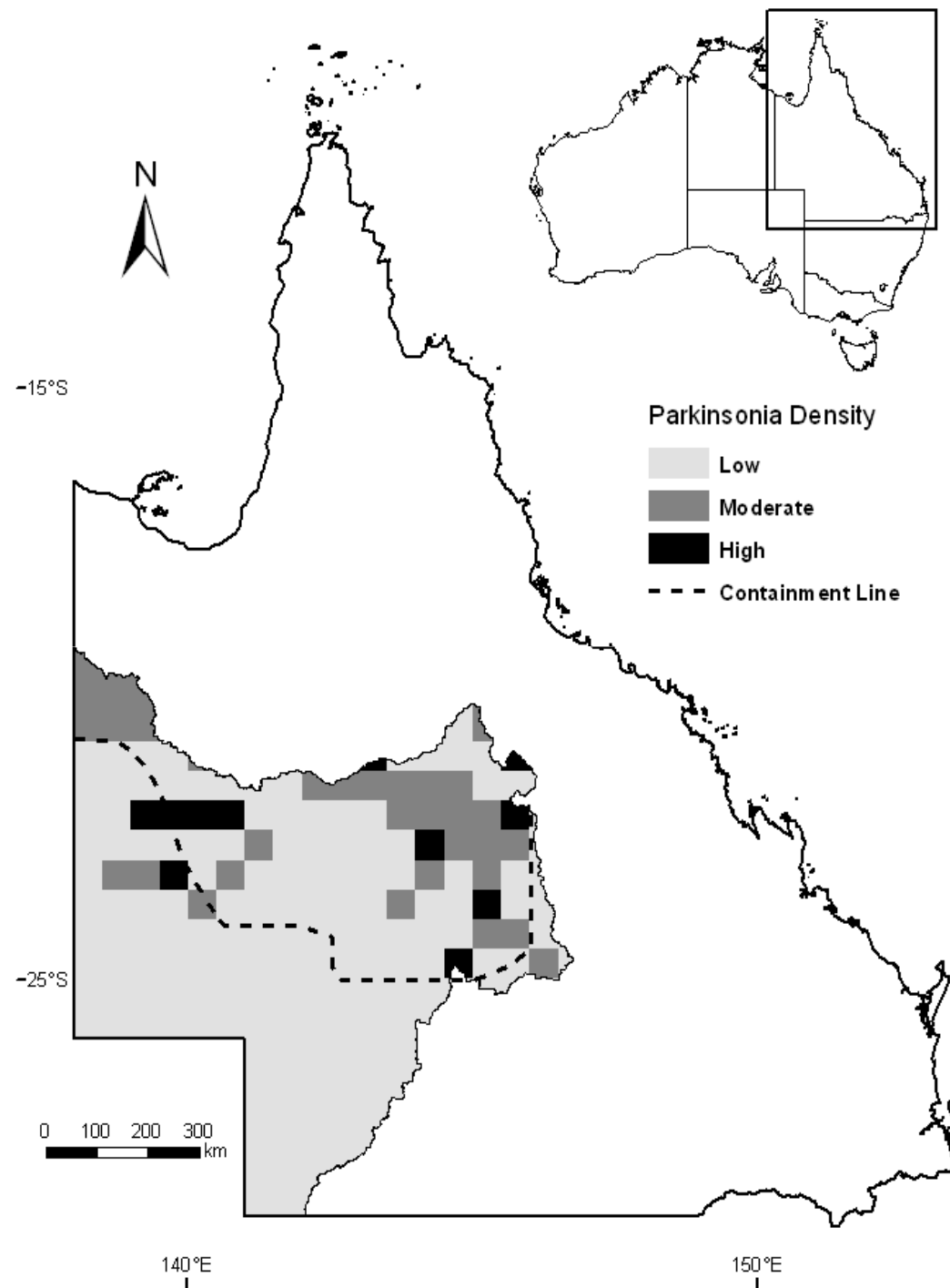


Susceptibility = High



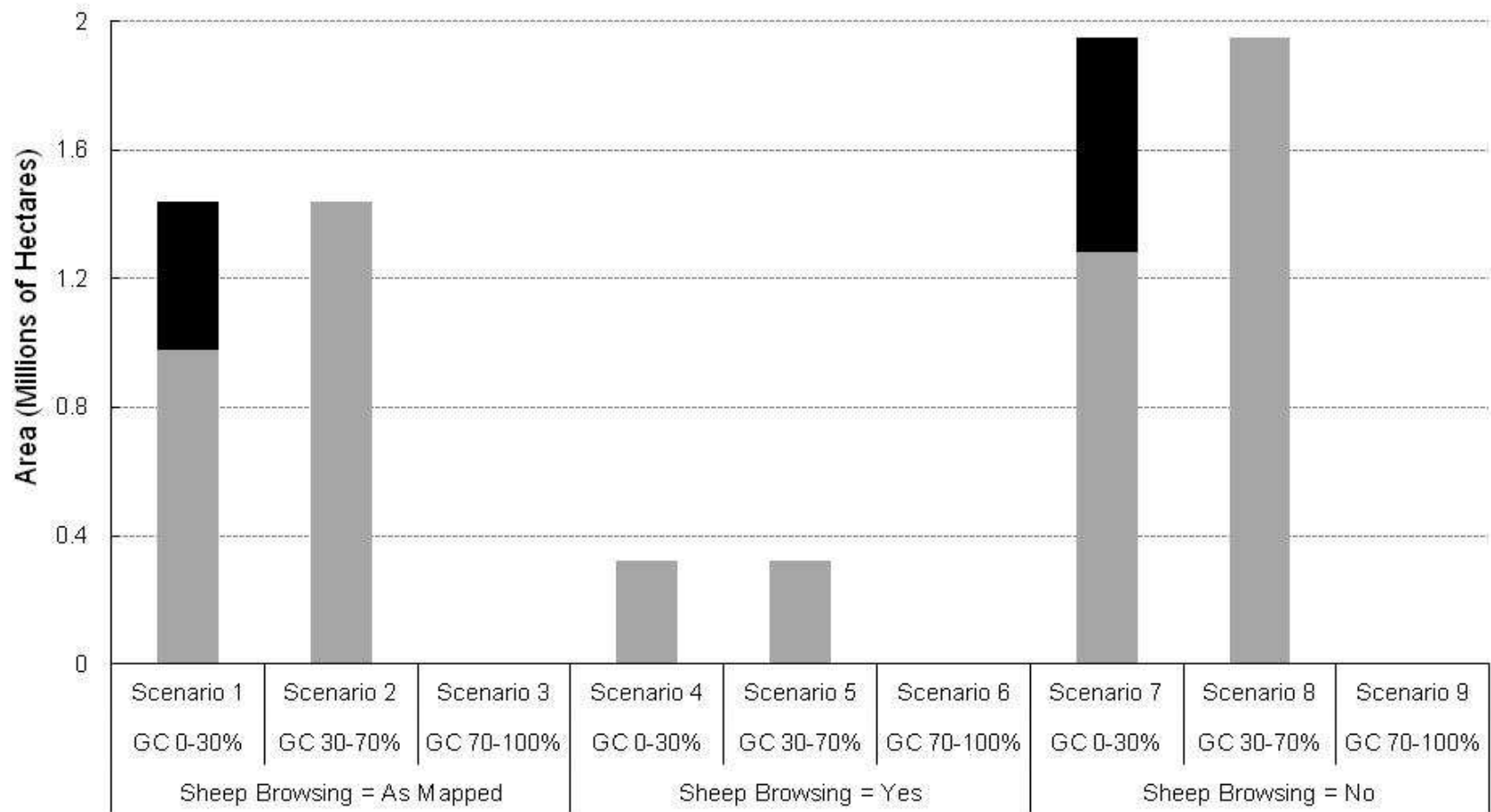
Susceptibility = High





Accuracy Assessment

Predicted Suitability	Estimated Density		Accuracy (%)
	Low	Moderate to High	
Low	54	8	87
Moderate to High	103	41	28
Accuracy (%)	34	84	



Area predicted as moderate (grey bars) and high (black bars) suitability for parkinsonia

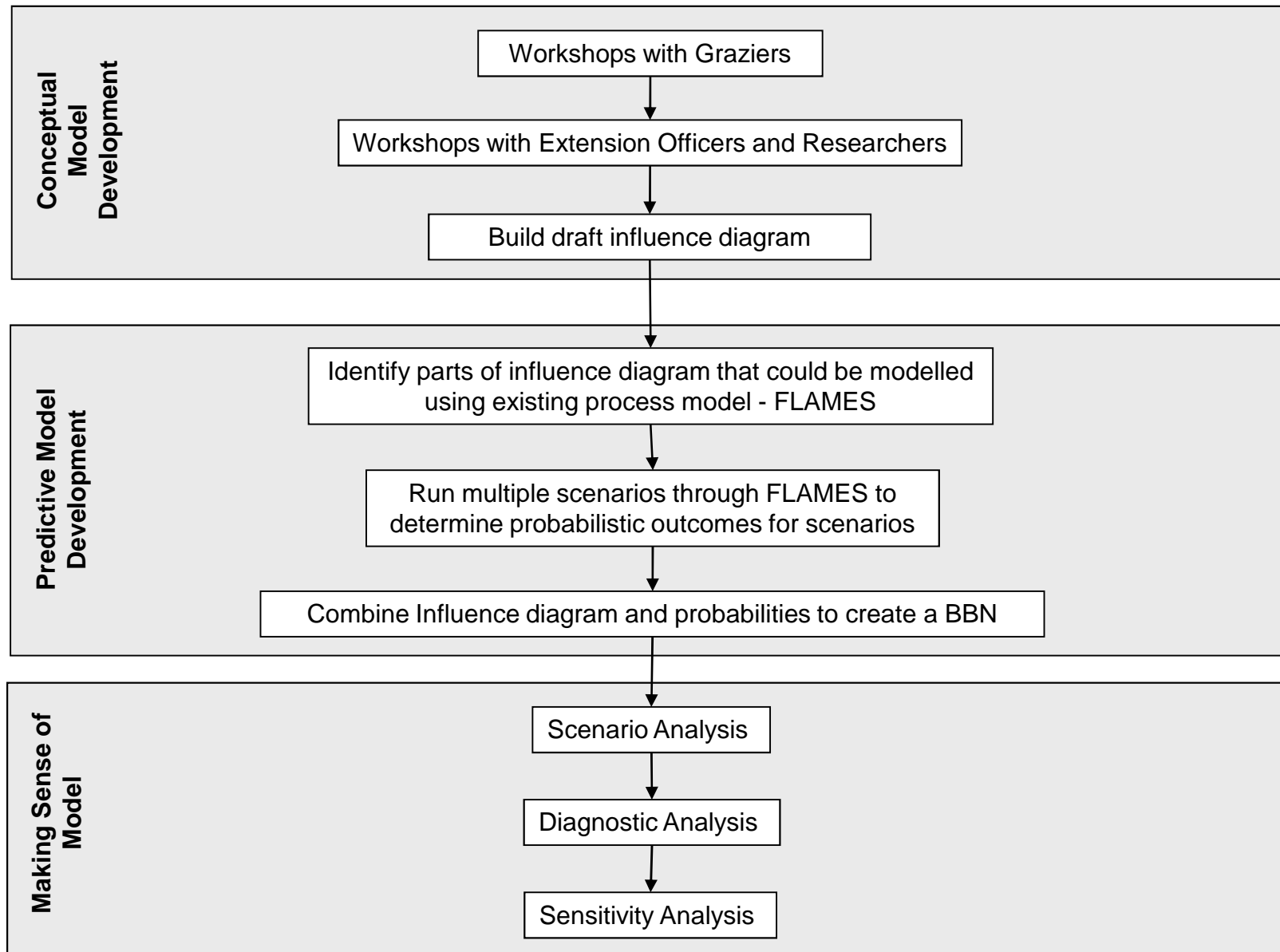
Example 4 – Tree Change in Australia

The Problem:

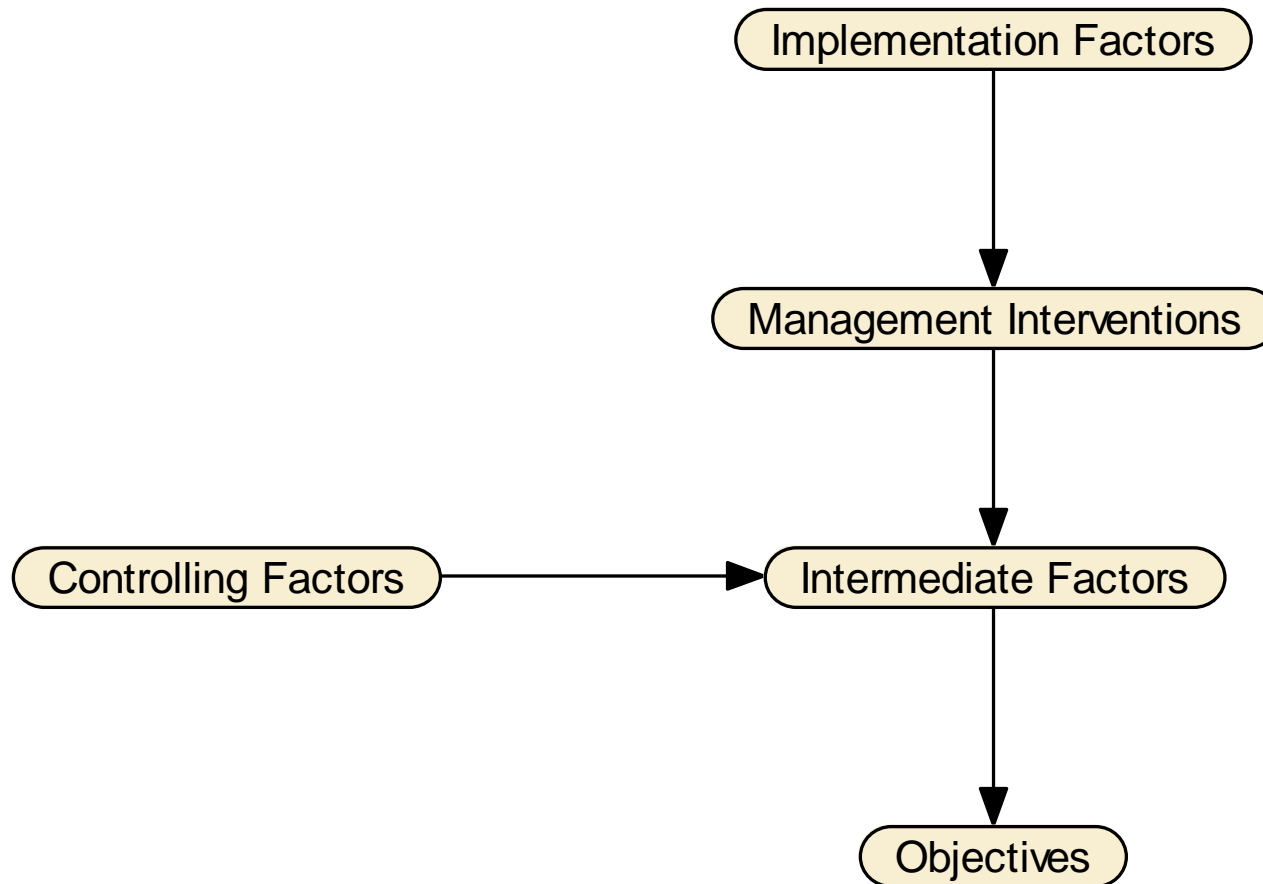
- Recently an increase in tree and shrub density has been reported in Australia's tropical savannas
- This has the potential to change catchment hydrology, carbon stocks, pasture biomass and wildlife habitat

What we did:

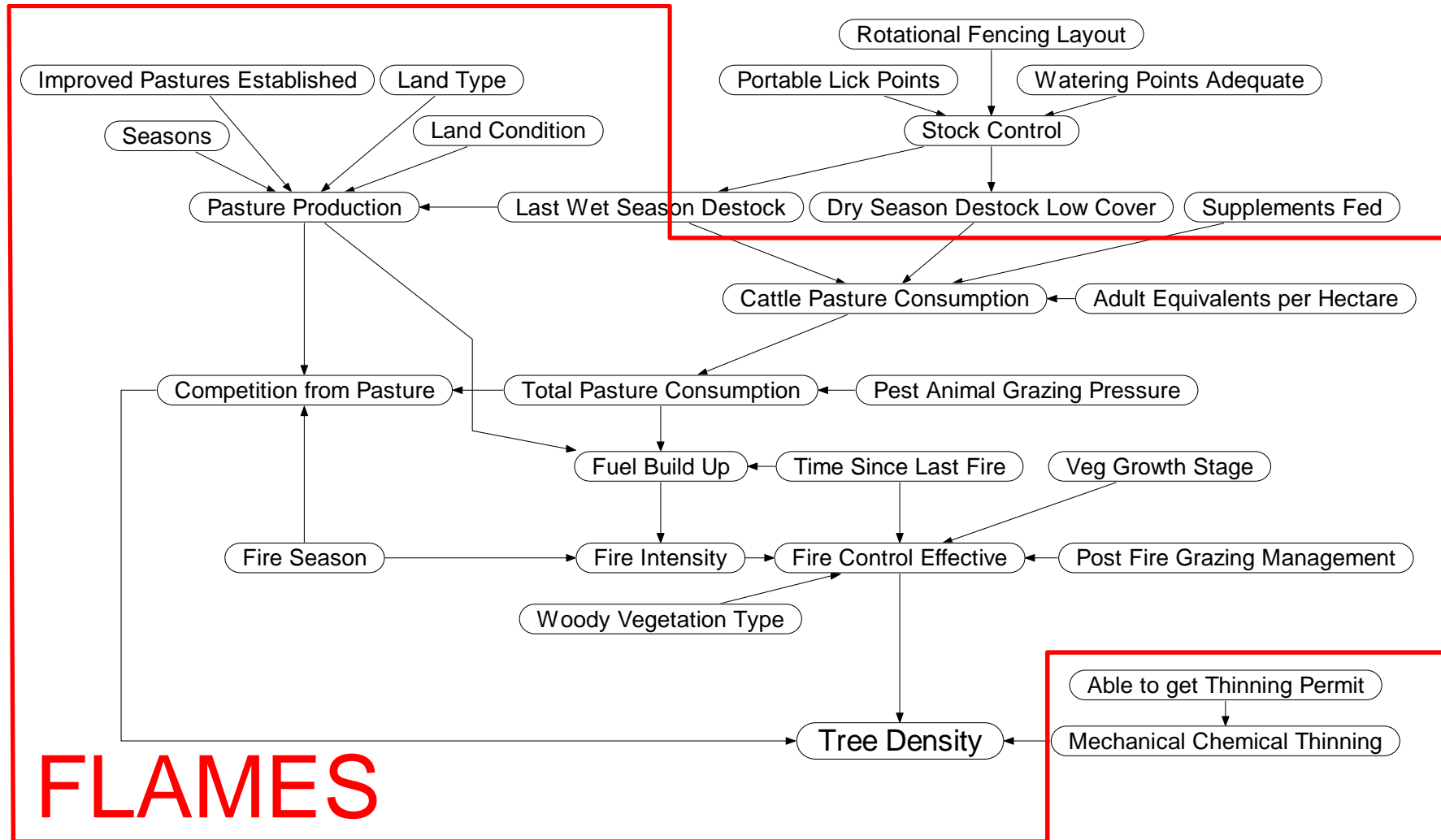
- Systems analysis with ecologists, extension officers and graziers



Grazier Workshops



Influence Diagram



Starting Stand

Year 0 360



After 10 Years

Year 10 334



After 20 Years

Year 20 290



After 30 Years

Year 30 261



After 40 Years

Year 40 230



After 50 Years

Year 50 192



After 60 Years

Year 60 168



After 70 Years

Year 70 137



© 2006 A.Liedloff, Tropical Savannas CRC, CSIRO

After 80 Years

Year 80 123



After 90 Years

Year 90 105

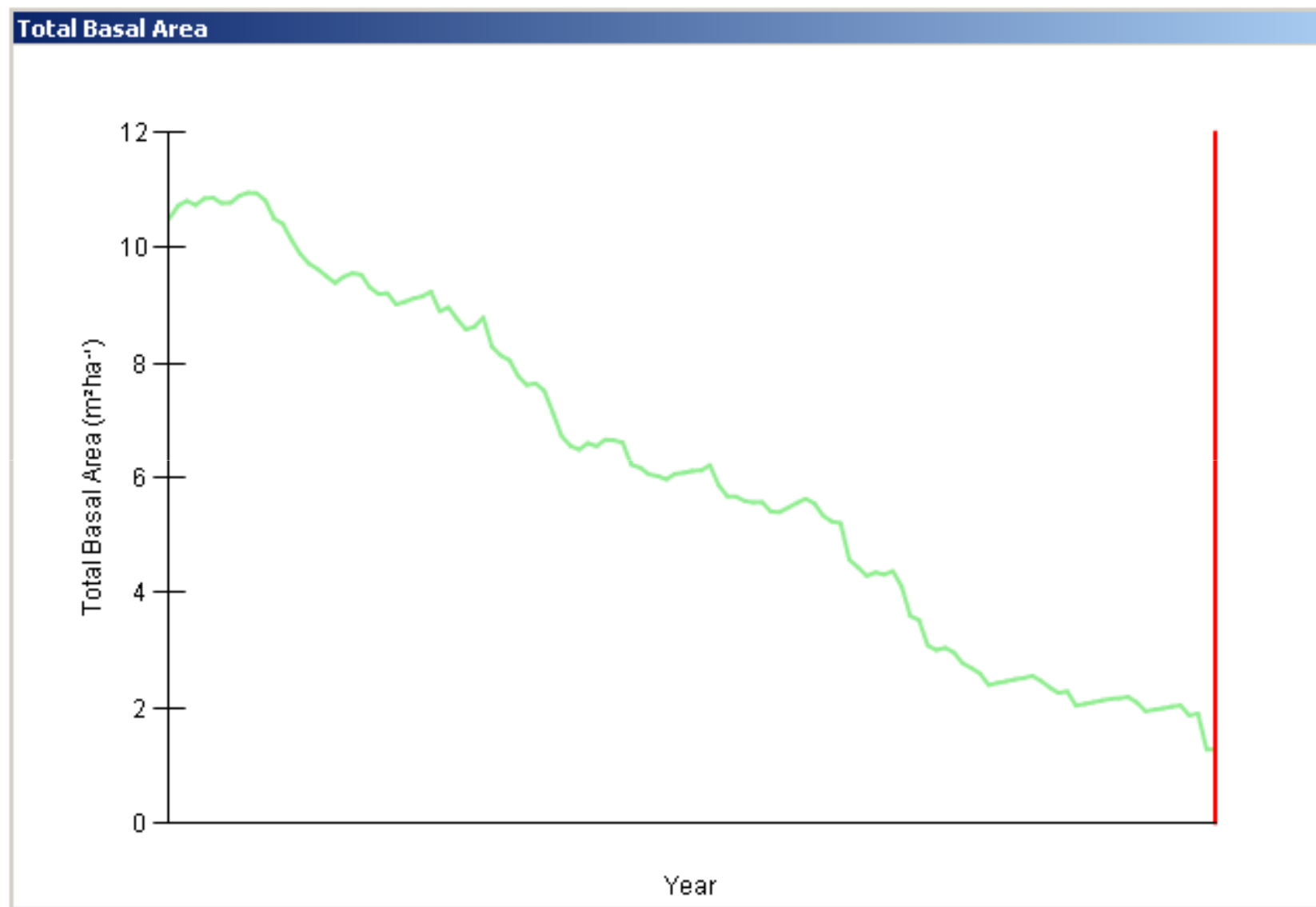


After 100 Years

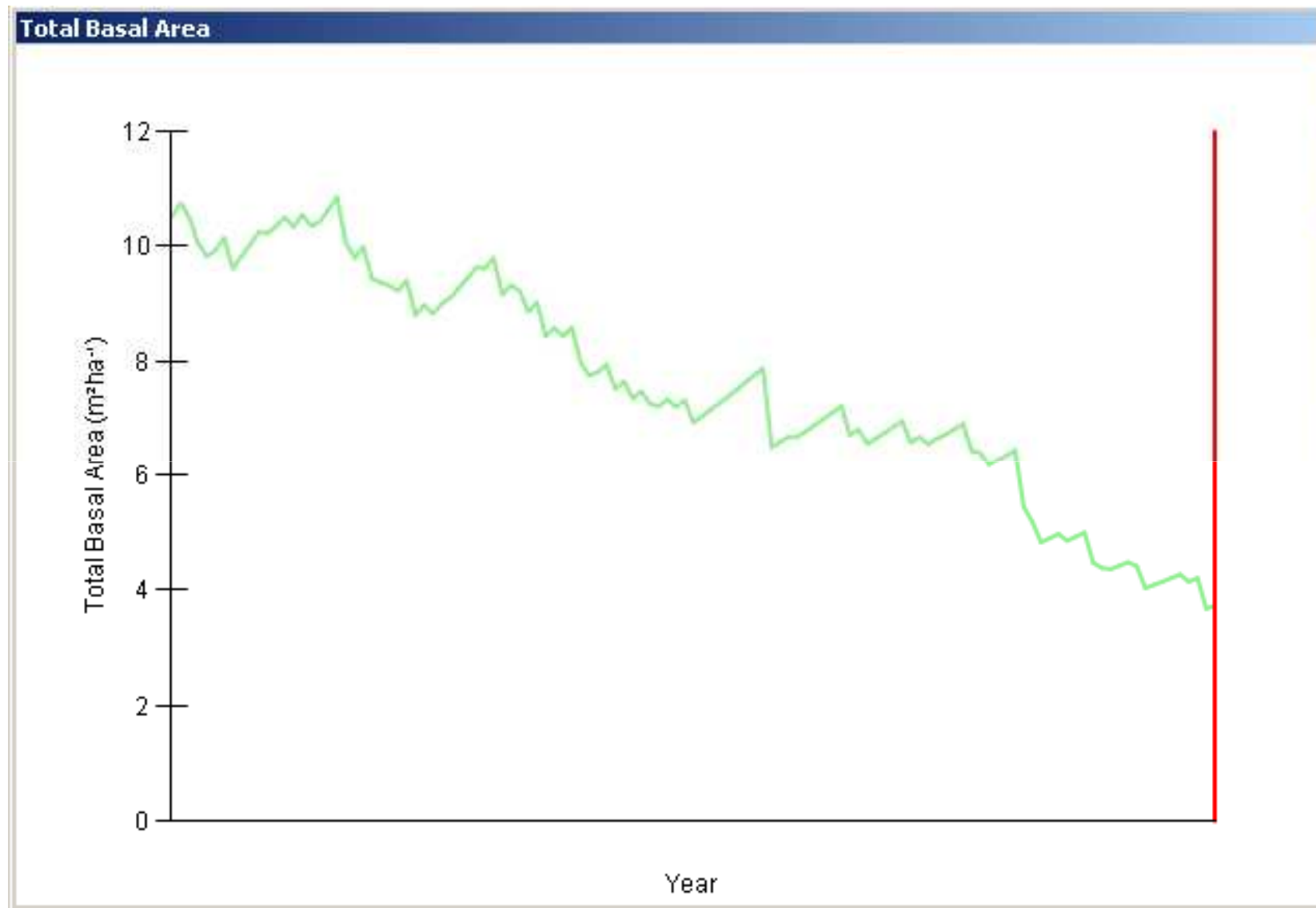
Year 100 91



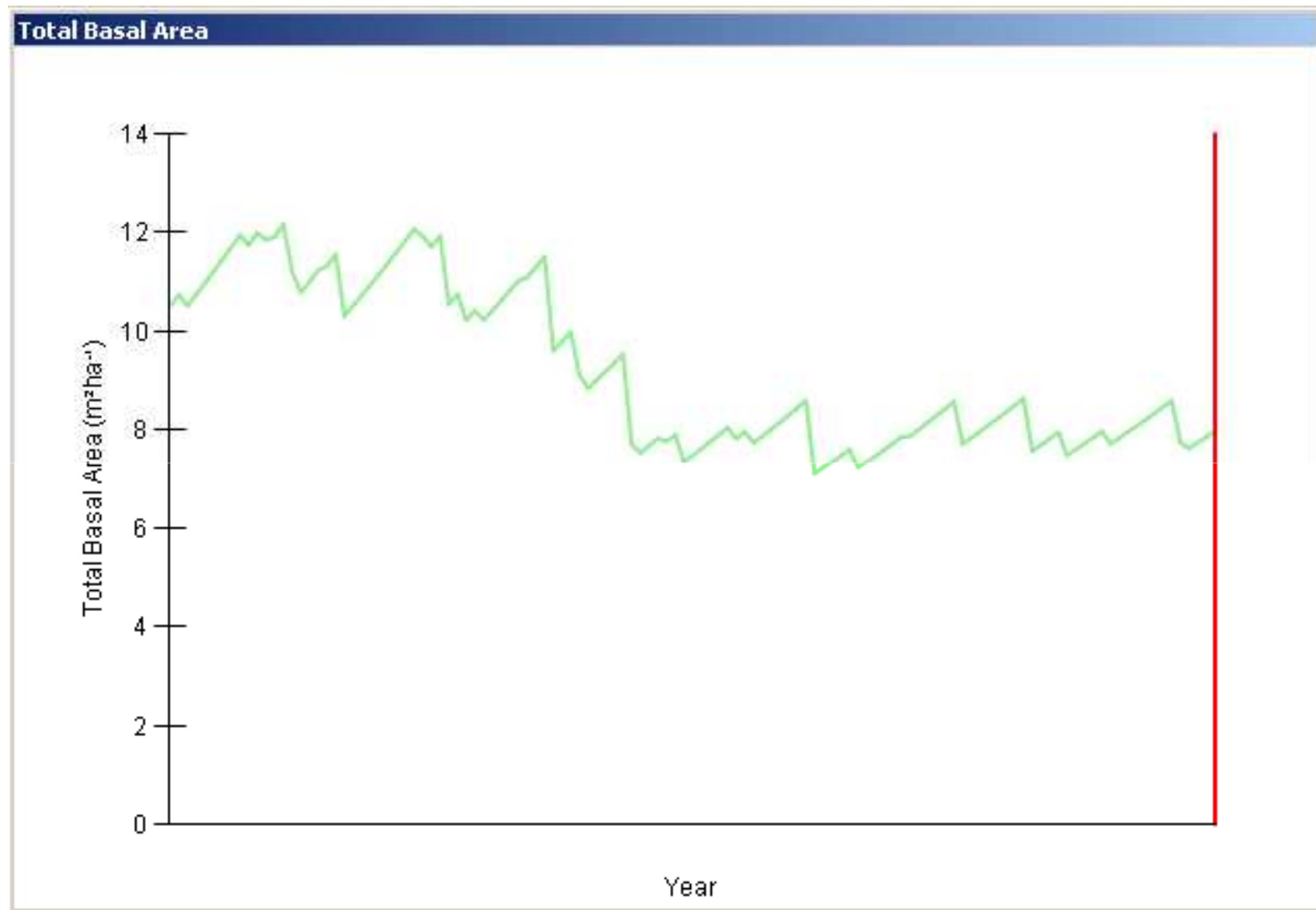
Annual Fires, Kakadu, No Grazing, Mature Staring Stand, *E. miniata*



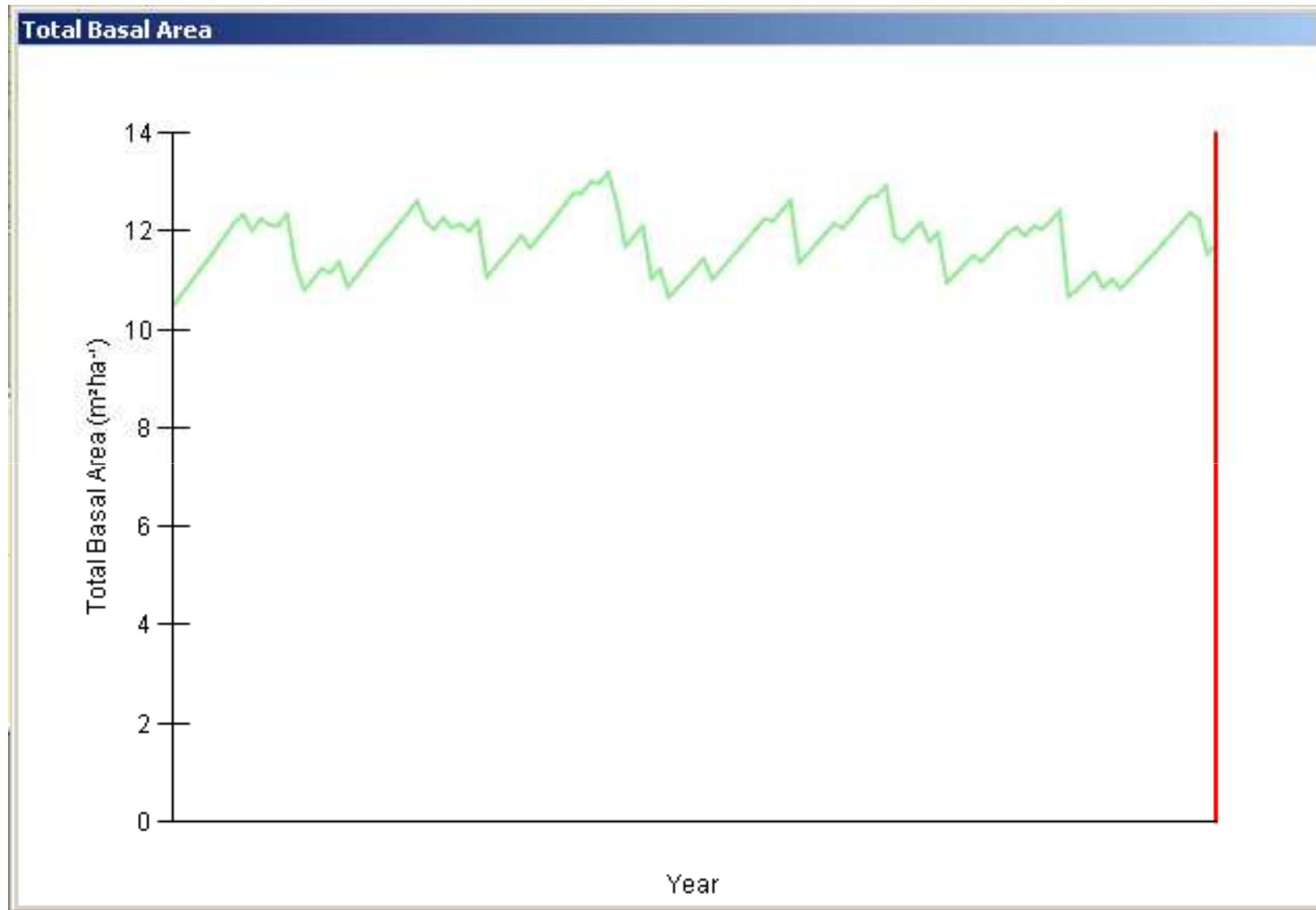
Biennial Fires, Kakadu, No Grazing, Mature Staring Stand, *E. miniata*

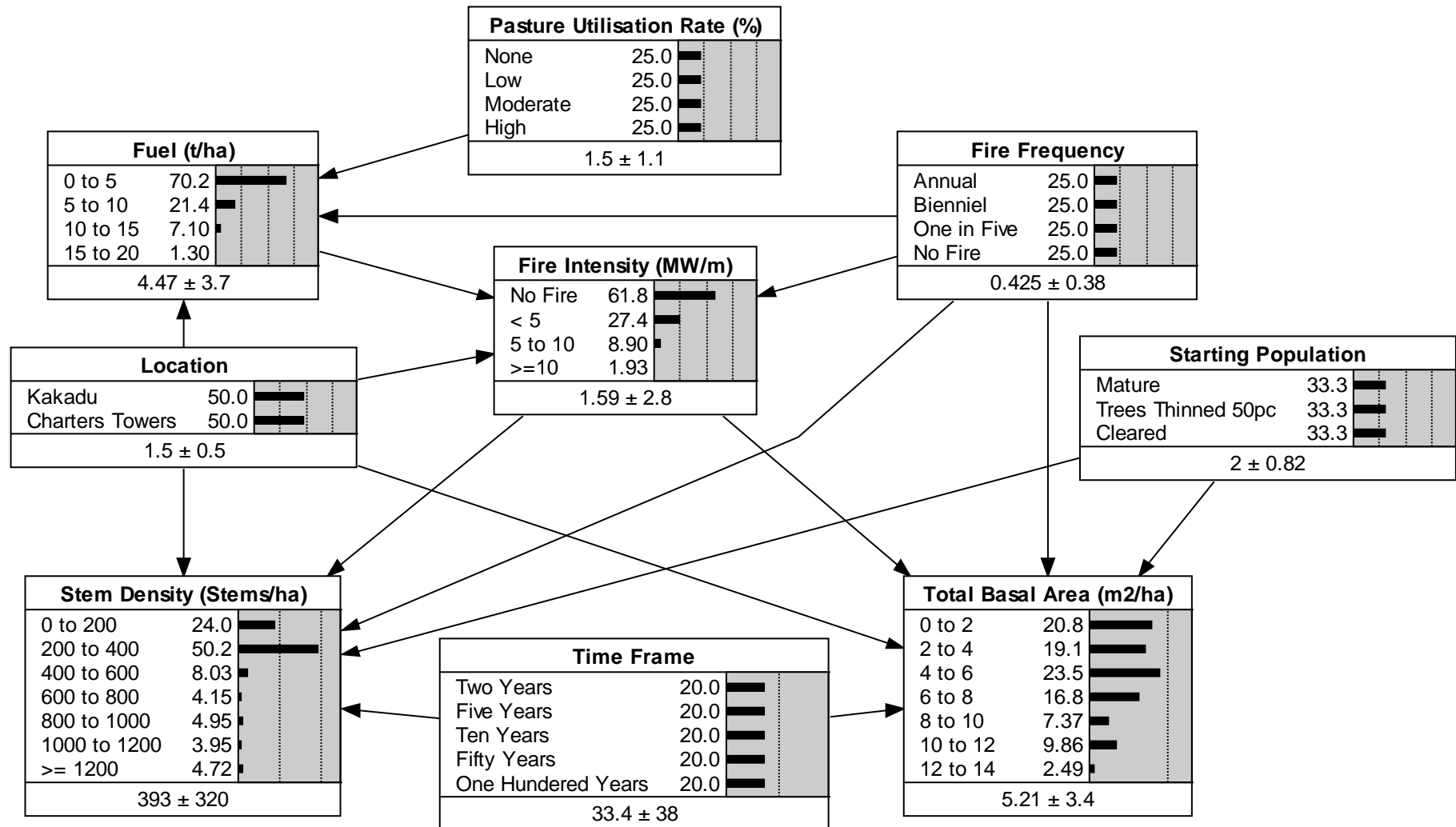


1 in 5 Year Fires, Kakadu, No Grazing, Mature Staring Stand, *E. miniata*



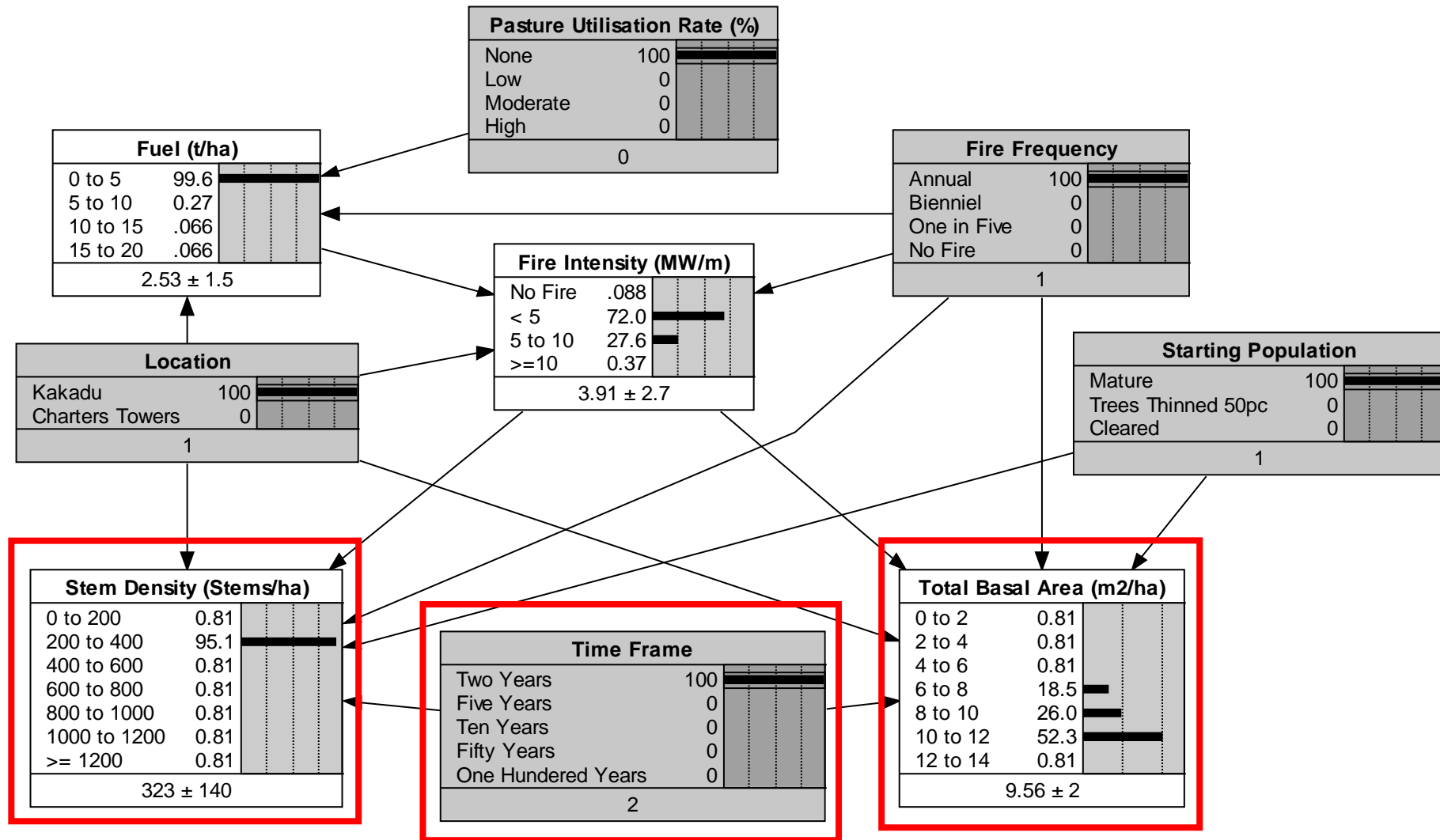
No Fire, Kakadu, No Grazing, Mature Staring Stand, *E. miniata*



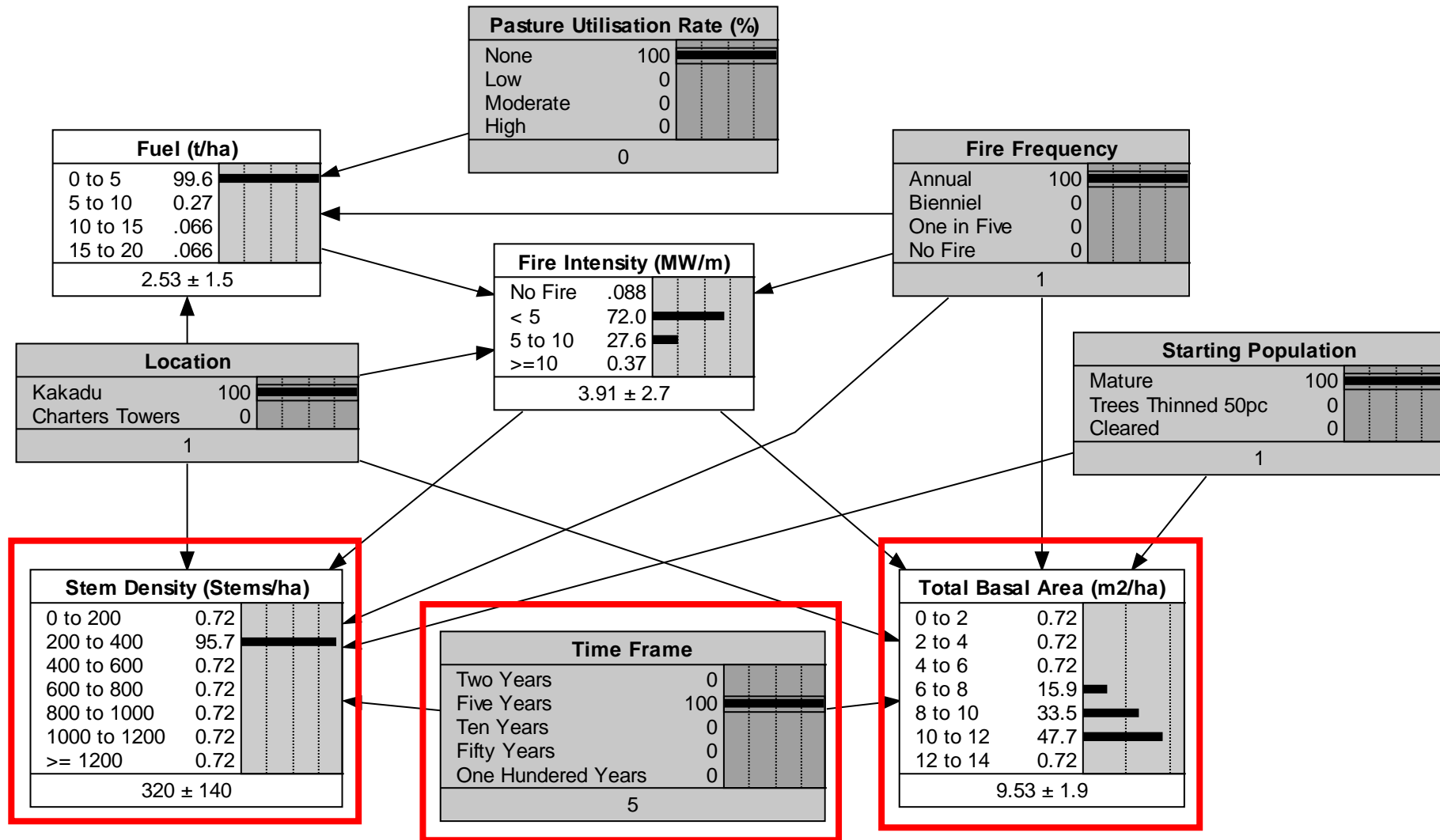


Annual Fire Frequency in Kakadu

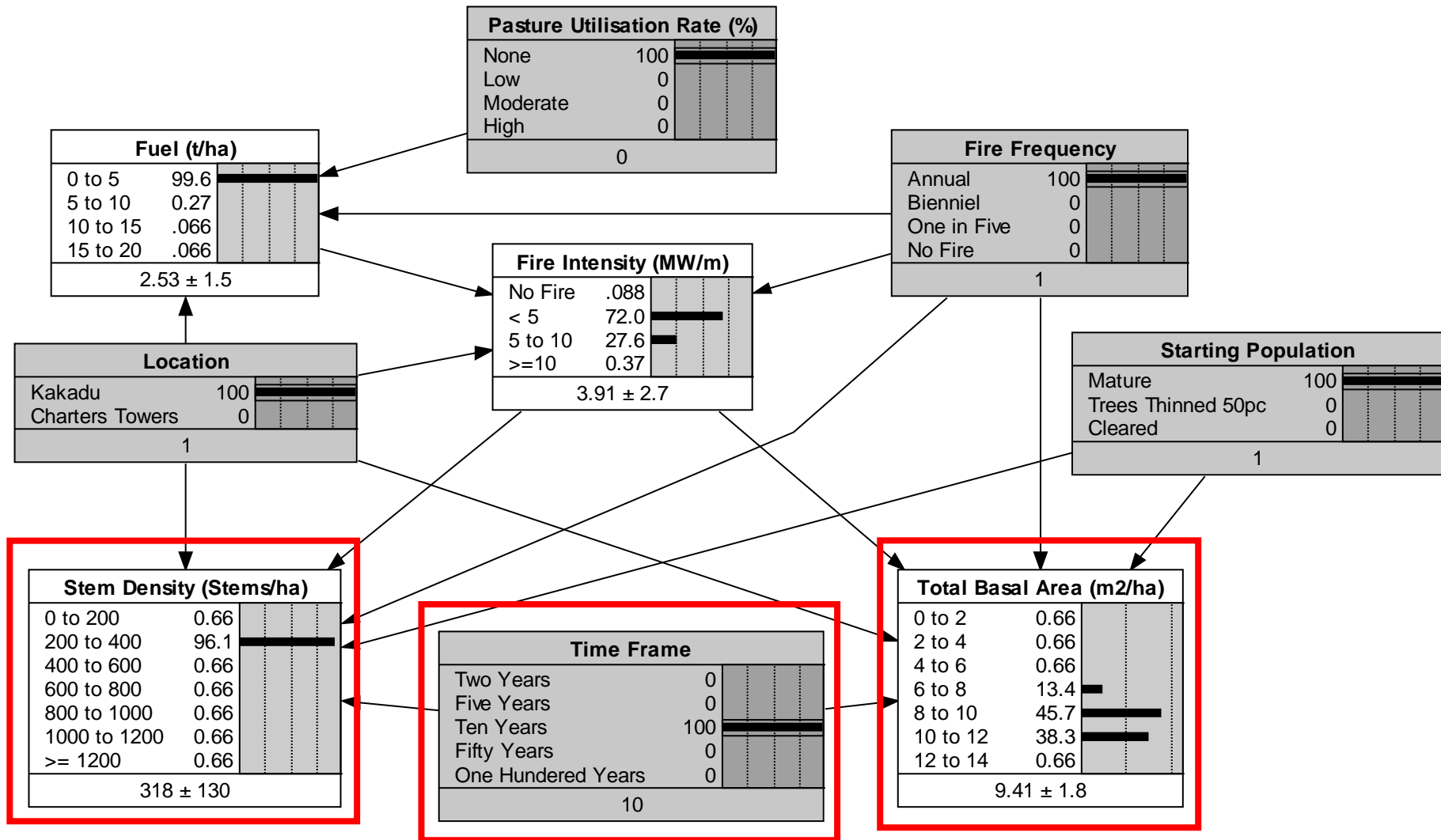
After 2 Years



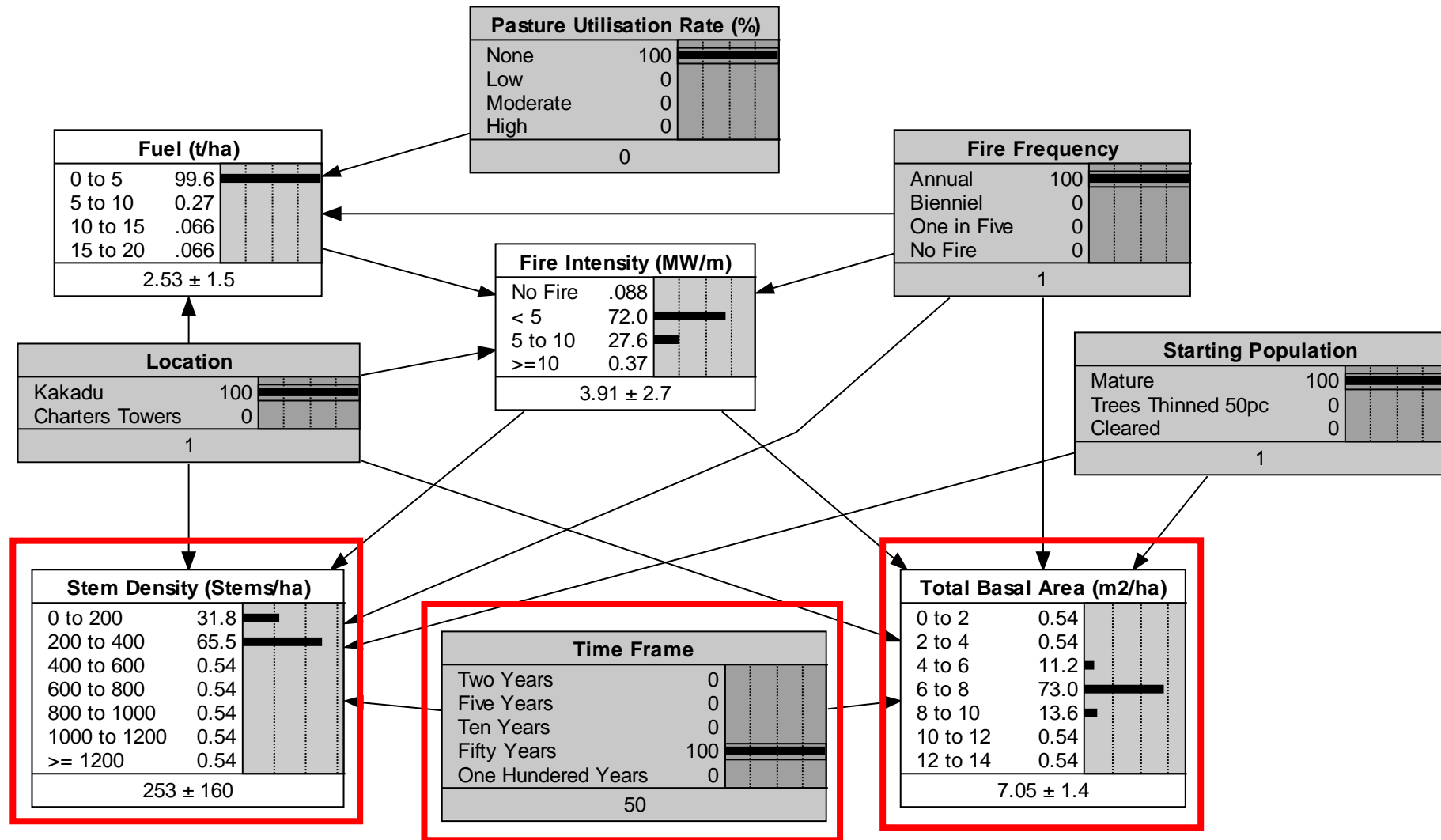
After 5 Years



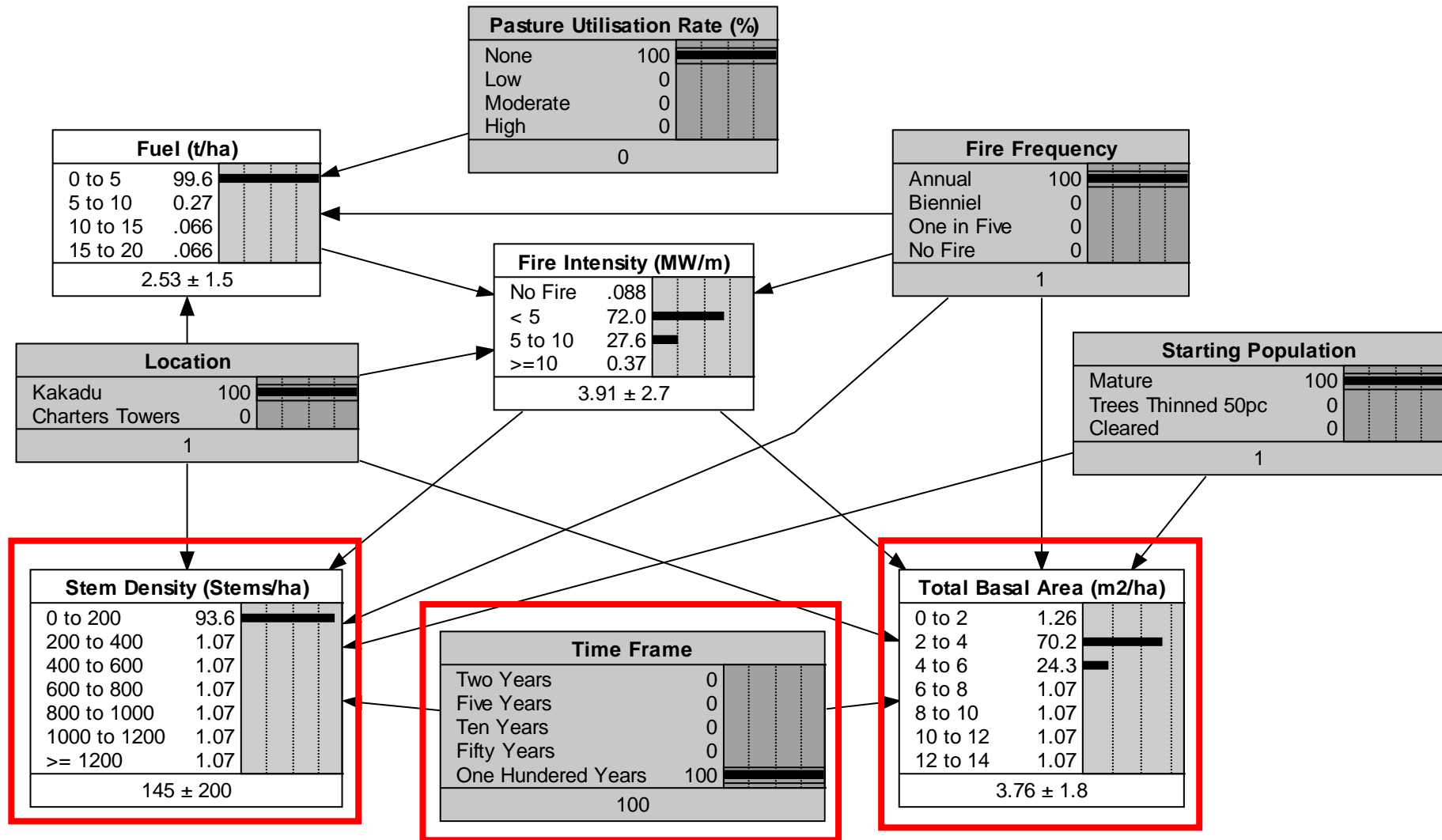
After 10 Years



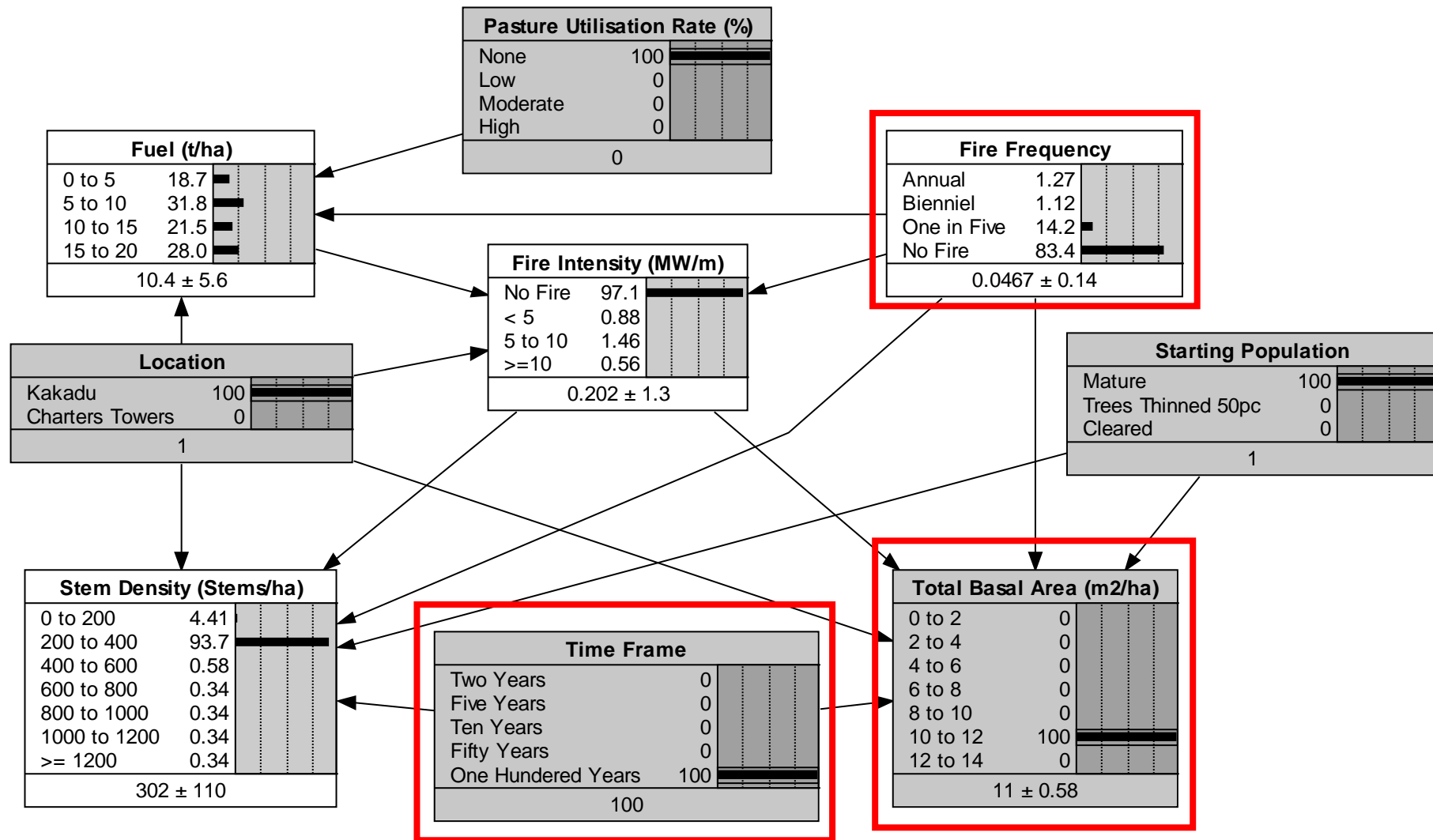
After 50 Years



After 100 Years

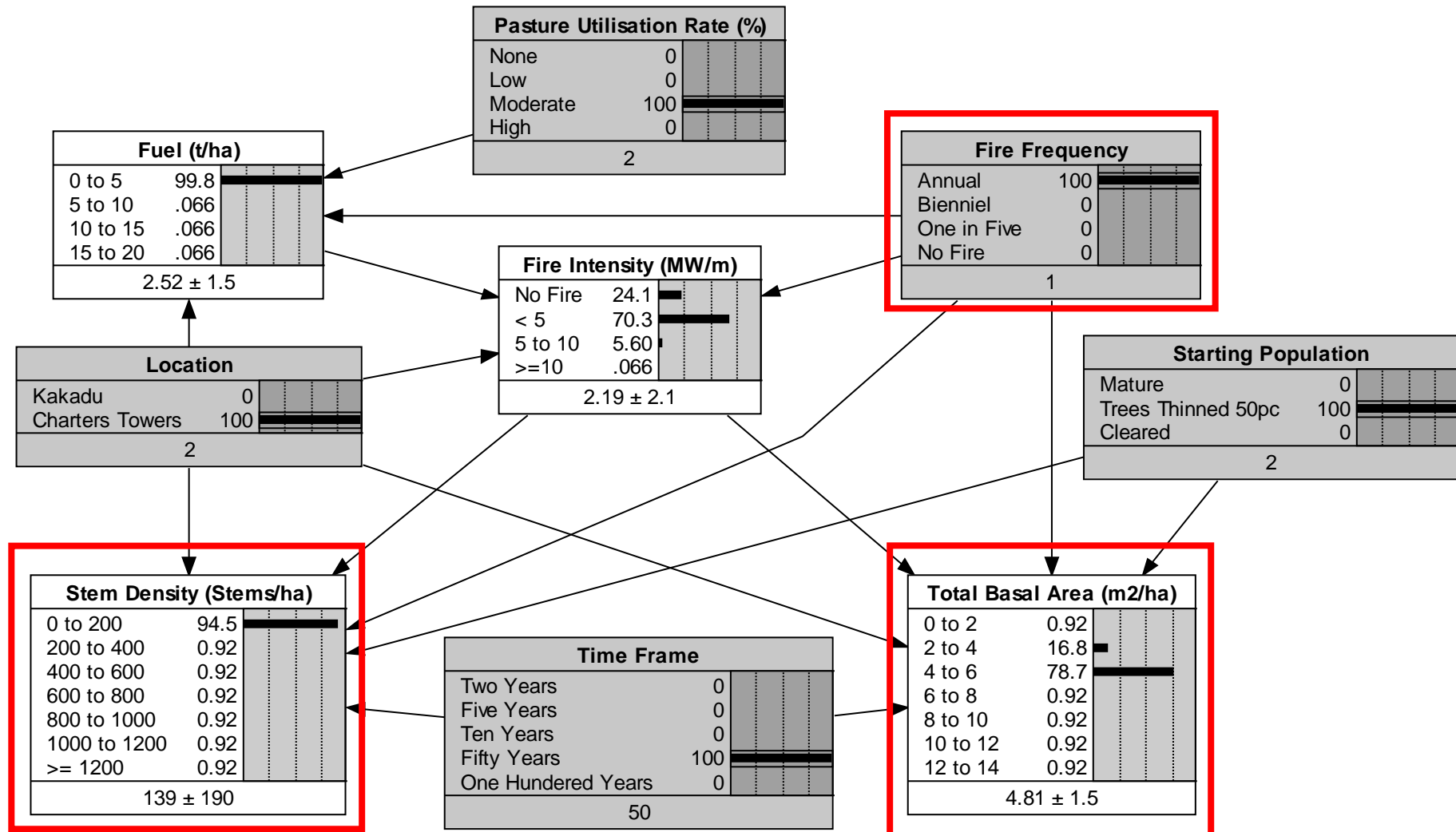


Diagnosing a fire frequency required to maintain a mature stand of trees with large TBA over a 100 year period in Kakadu National Park.

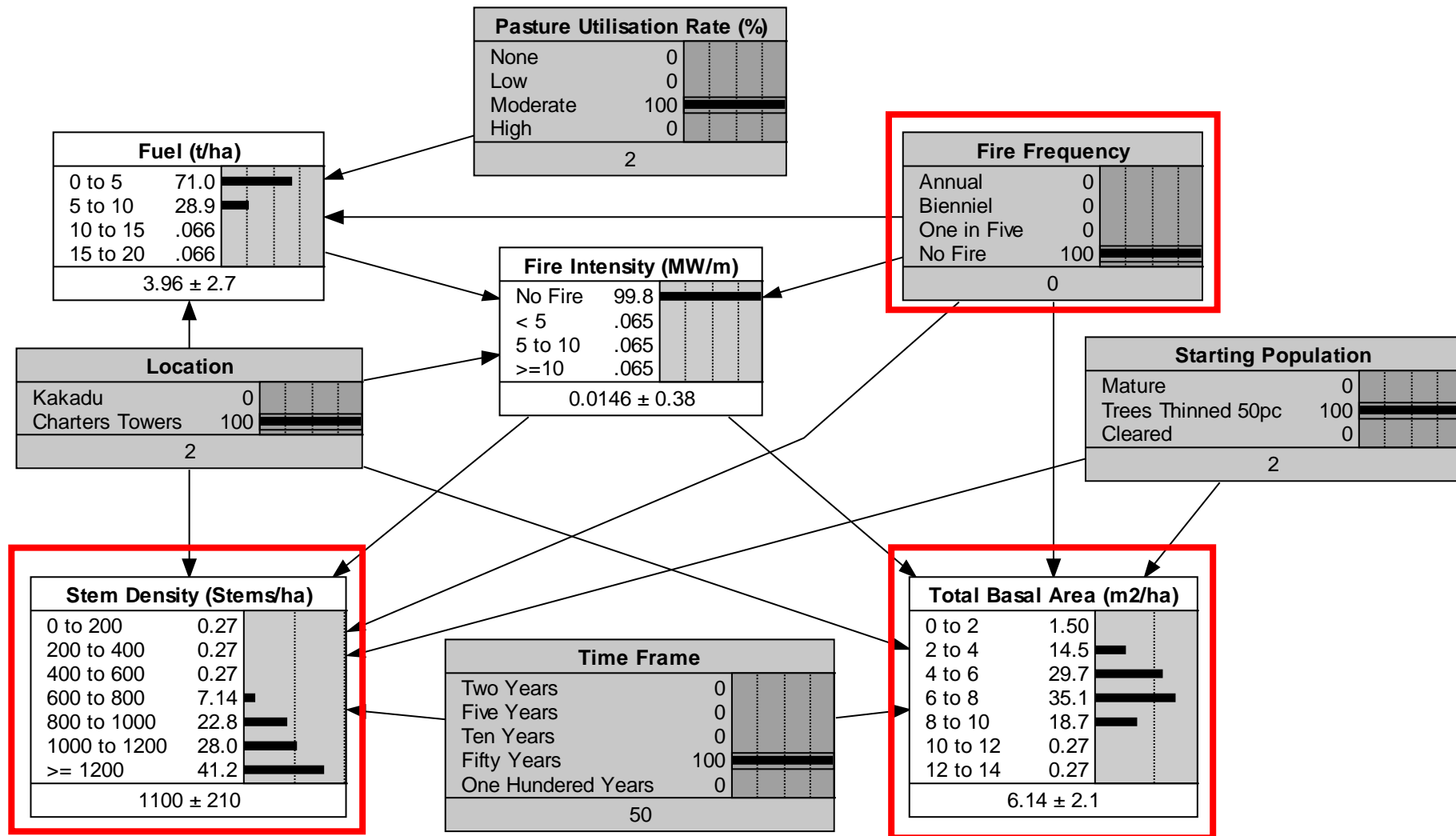


Predicting the effect of burning on stem density and TBA of a thinned tree stand after fifty years with moderate grazing pressure in Charters Towers.

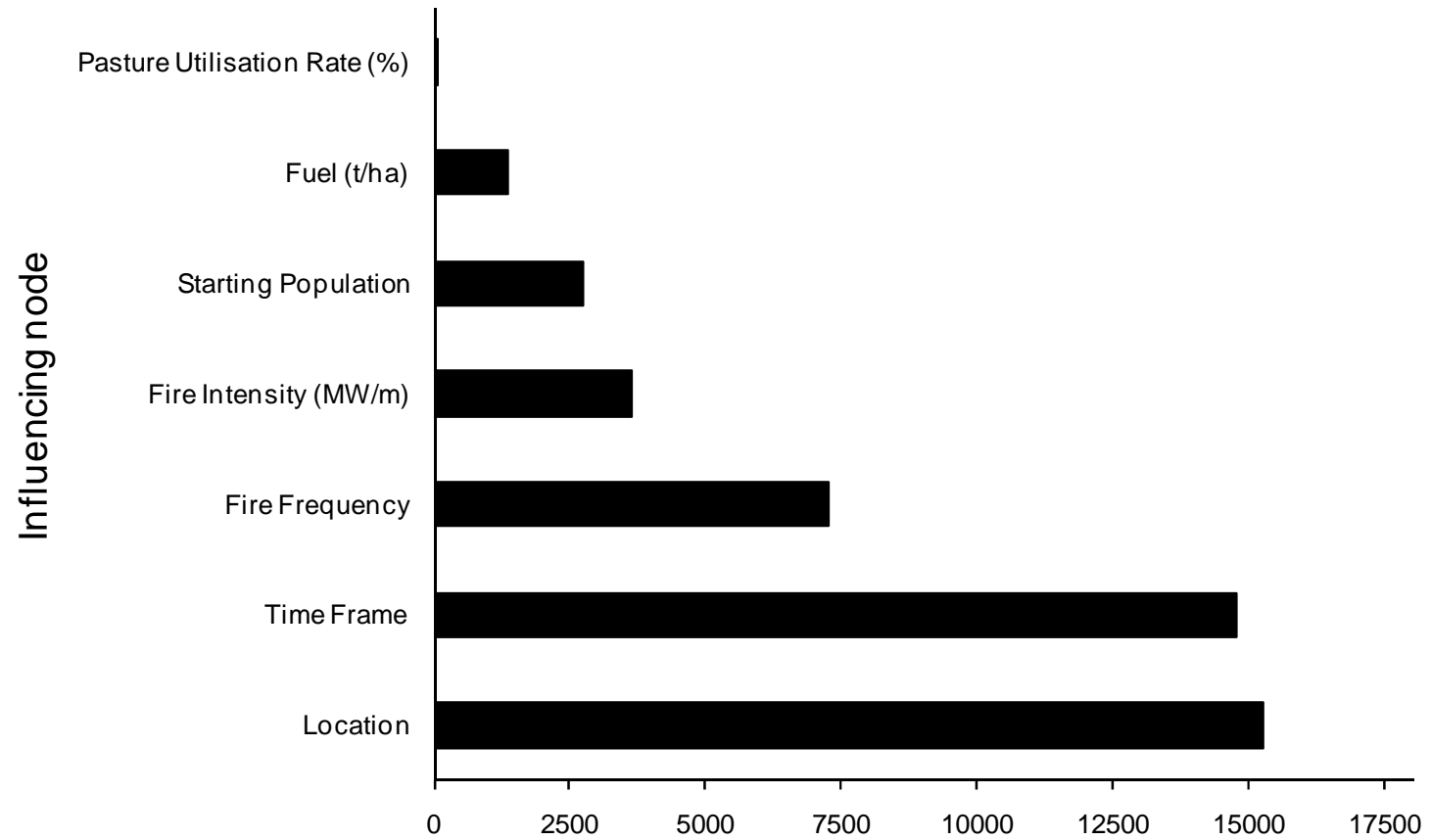
Annual Fires



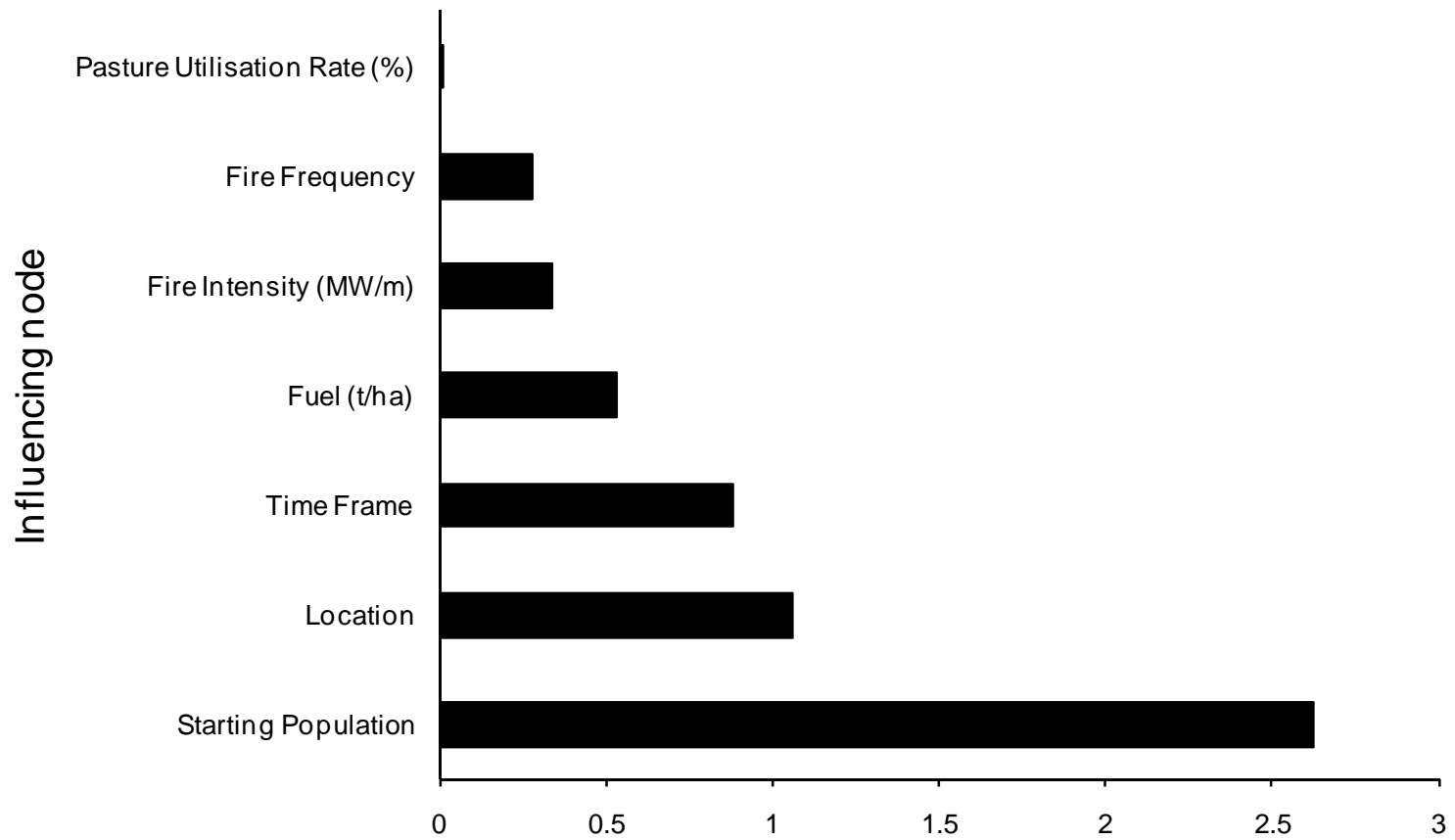
No Fire



Sensitivity Analysis - Stems per Hectare



Sensitivity Analysis – Total Basal Area



Conclusions

- In the short term (after 2 years), tree numbers are most influenced by the Starting Point (Mature, Thinned or Cleared stand) followed by Fire Intensity and Fuel
- In the long term (after 100 years), tree numbers are most influenced by Location followed by Fire Frequency and Fire Intensity (assuming the woody vegetation type is Eucalypt)
- This shows that in the long term, climate is the factor driving tree numbers

Take Home Messages

- Systems thinking - mechanism for knowledge integration, stakeholder participation and intervention design
- Bayesian Networks a flexible tool for use in participatory systems analysis – scenario, sensitivity and diagnostic analysis

Take Home Messages

- Embedding these within the Llab cycle – the interventions/actions are implemented and then reviewed (step 7 – reflection)

