

# Learning about colonisation under an adaptive management framework

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Supervisors:

Mick McCarthy, Brendan Wintle, Eve McDonald-Madden



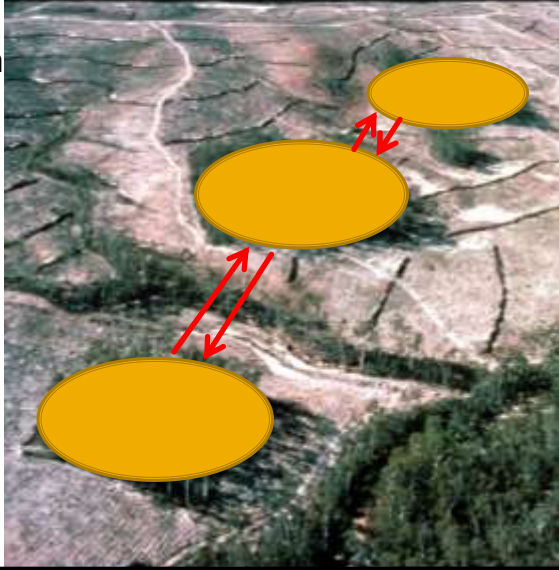
## Habitat fragmentation

- Habitat fragmentation a major threat to biodiversity
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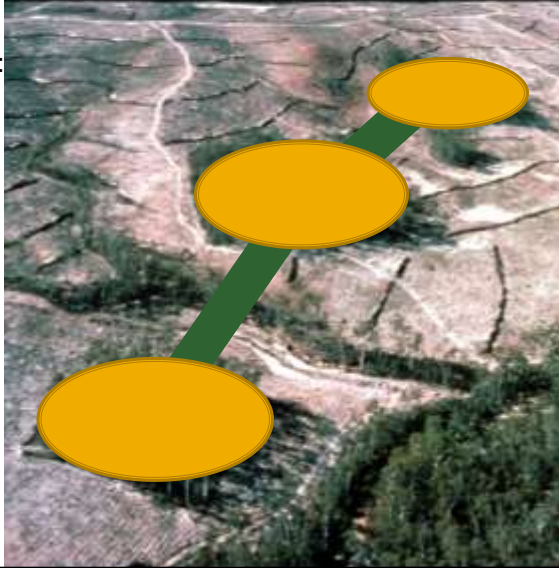
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## Adaptive management

- Adaptive management is a framework for learning while doing
  - Passive adaptive management
  - Active adaptive management
- Applied to many areas in conservation science
- Yet to be applied to meta-population restoration



## Aim of research

- Find the optimal management strategy for a meta-population
- Research questions:
  1. *What is the optimal management strategy for a meta-population when colonisation is uncertain?*
  2. *What is the optimal way to learn about colonisation as we manage a meta-population?*

## Case study: Growling Grass Frog



- When do we improve existing patches?
- When do we add patches to the meta-population?

## Method

- 1. Meta-population model
- 2. Incorporate uncertainty into the model
- 3. Uncertainty updated during modelling
- 4. Find optimal management and learning strategy

## Meta-population model

- Adapted from Nicol and Possingham (2010)
- Number of patches  $\mathbf{M} \in \{M_{min}, M_{min+1}, \dots, M_{max-1}, M_{max}\}$
- Area of patches  $\mathbf{A} \in \{A_{min}, A_{min} + k, \dots, A_{max} - k, A_{max}\}$
- Number of occupied patches  $\mathbf{n} \in \{0, 1, 2, \dots, M\}$
- What is the probability of  $\mathbf{M}$ ,  $\mathbf{A}$  and  $\mathbf{n}$  changing in the next time step?

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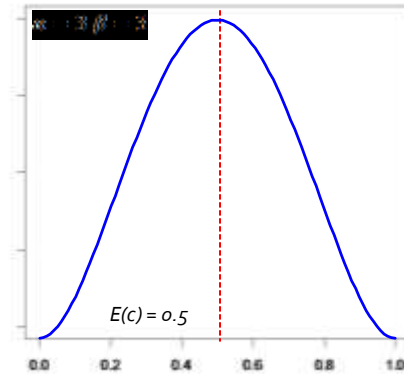
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### Management actions

- 1. Add area to existing patches
- 2. Create new patches

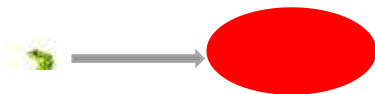
## Incorporating uncertainty in colonisation

- $c$  modelled with Beta distribution
- Shape defined by parameters: alpha and beta
- Expected value of colonisation equal to mean

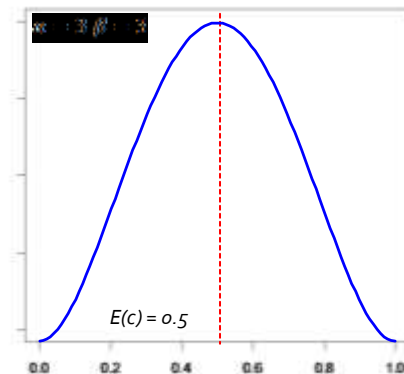


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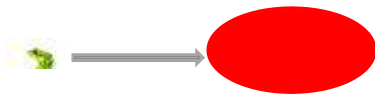
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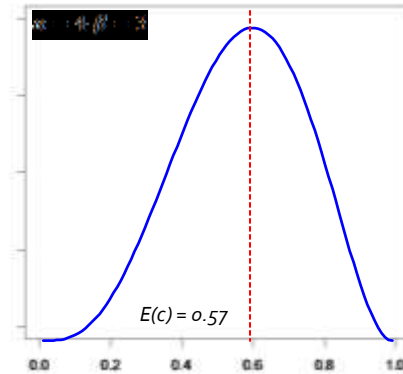


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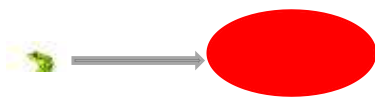


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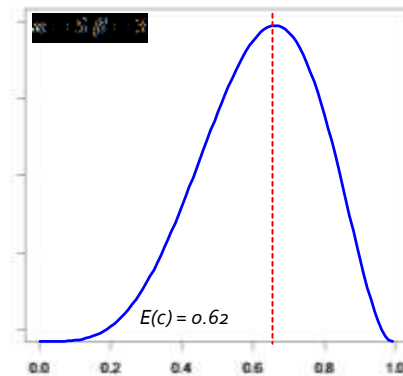


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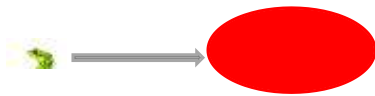


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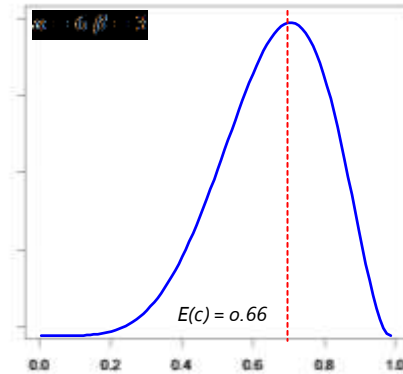


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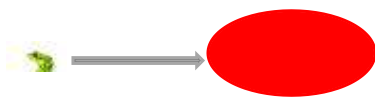


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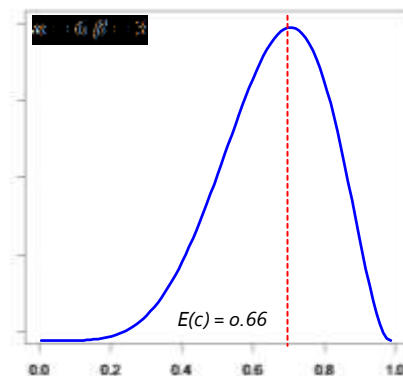


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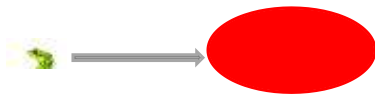


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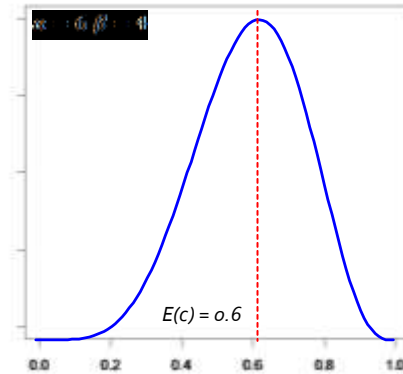


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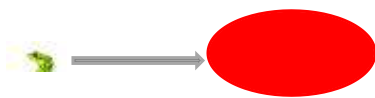


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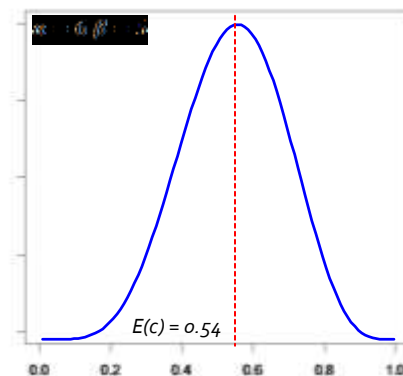


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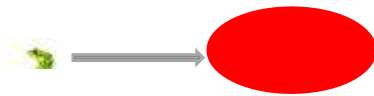


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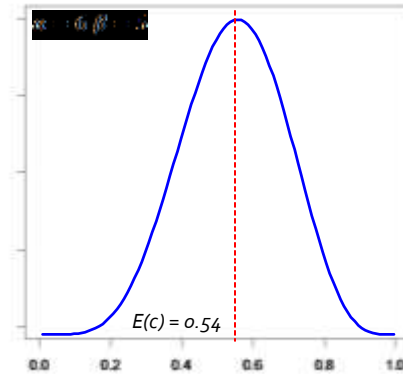


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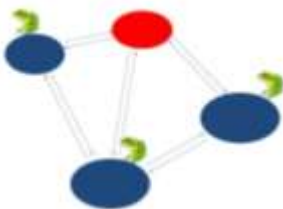


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## Learning while managing

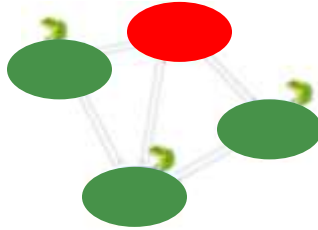
Do nothing



- 1 learning opportunity
- No return

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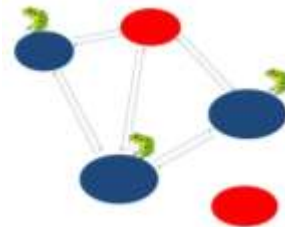
Add area



- 1 learning opportunity
- Extinction rate reduced

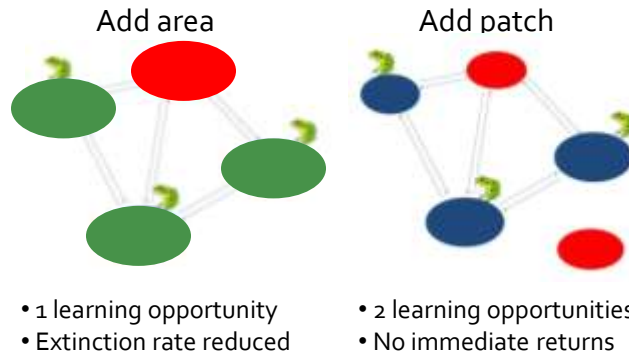
## Learning while managing

Add patch



- 2 learning opportunities
- No immediate returns

## Learning while managing



- Which action should we implement given our belief in colonisation?
- Which action should we implement to manage and learn?

## Optimisation

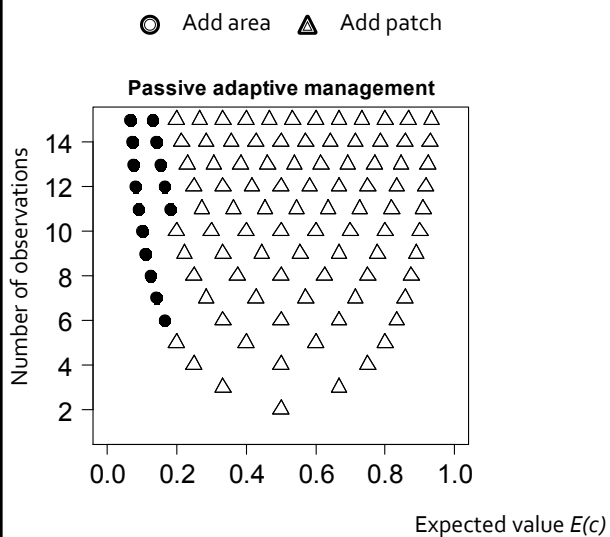
- Optimal strategy found using **stochastic dynamic programming (SDP)**
- Steps of the SDP:
  - 1. Set time horizon and management objective
  - 2. Loops through each time step
  - 3. Searches through each state – all combinations of  $M$ ,  $A$ ,  $n$ ,  $\alpha$  and  $\beta$
  - 4. Finds optimal action using the **meta-population model**

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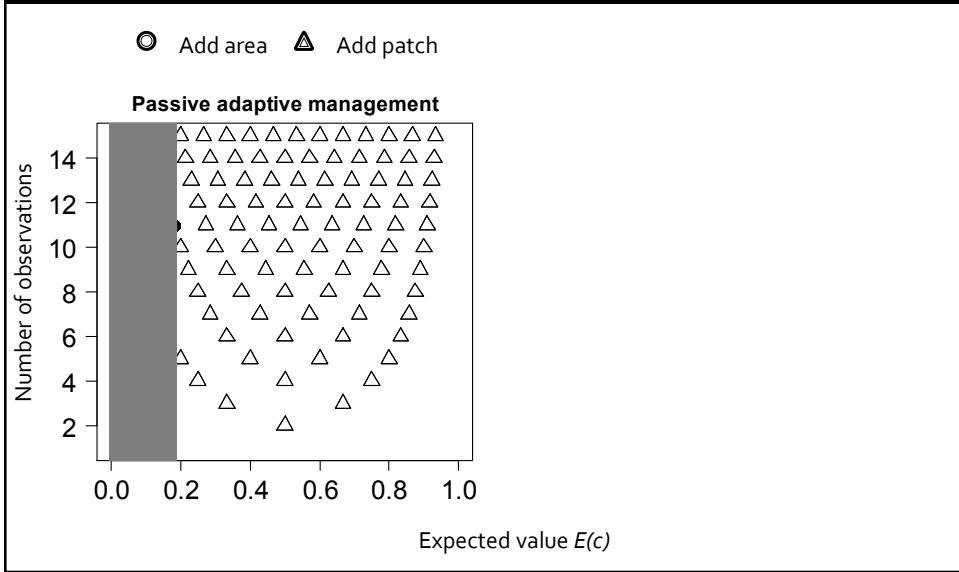
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Updates **belief in colonisation** over time
- Passive adaptive management
  - Beliefs updated retrospectively
- Active adaptive management
  - Value of learning considered prior to management

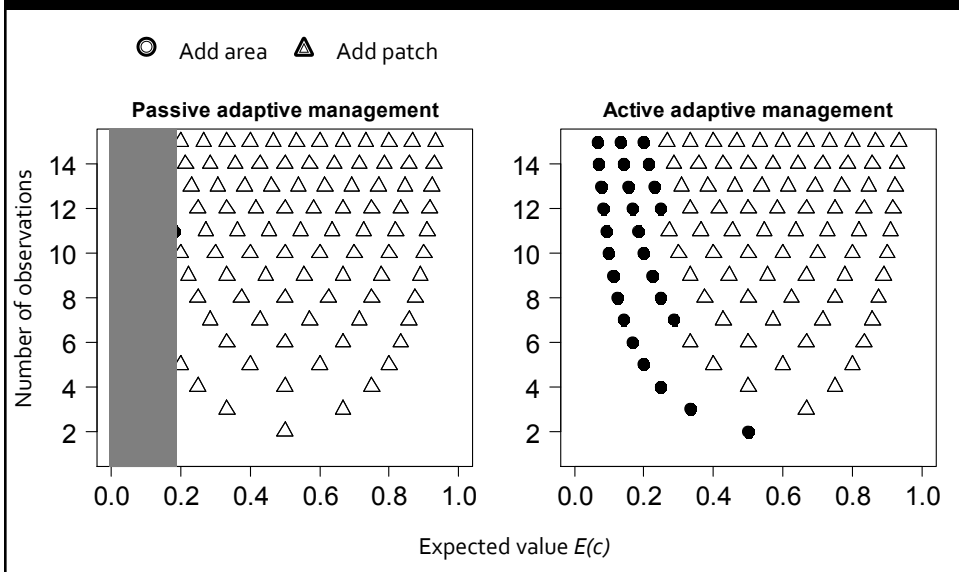
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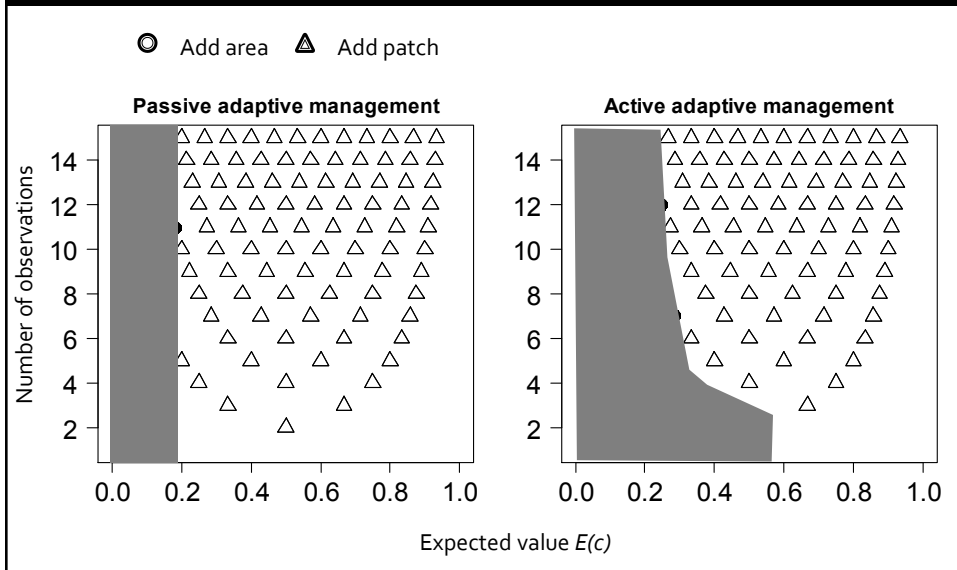


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## Discussion

- Meta-population assumed to be spatially implicit
- Colonisation not a function of the number of occupied patches
- Number of patches occupied,  $n$ , known with certainty
- Relationship between patch area and extinction known with certainty

## Summary

- Many species will occupy meta-populations
- Models are useful tools for guiding meta-population restoration
- Model parameters often highly uncertain
- Adaptive management provides a way to act immediately and resolve key uncertainty
- Increasingly important in conservation science

## References

- Nicol S.C., and Possingham H (2010) Should metapopulation strategies increase patch area or number of patches? Ecological applications 20(2) :566-581