

Optimal Selection of Habitat Reserves

Bala Krishnamoorthy

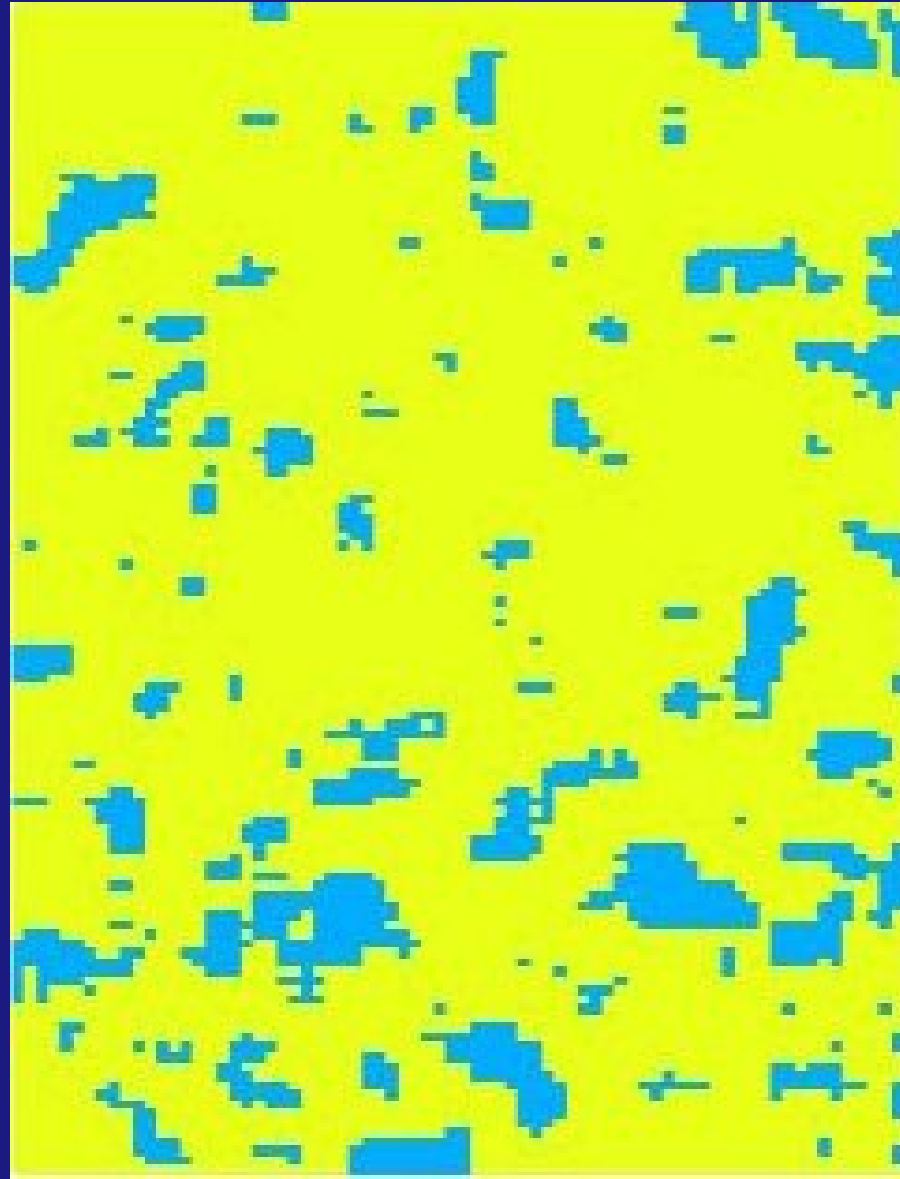
joint work with
Natalie Baerlocher, David Allen
NSF-UBM Project

INFORMS 2007 Seattle

The Problem

- Habitat reserve design for territorial disperser
e.g., the Northern Spotted Owl (NSO)
- Binary Habitat
Green = Suitable
Beige = Unsuitable
- Declining population
- Which habitat areas will best augment an existing reserve area?

NSO Habitat Landscape



Life Cycle Parameters

- Territorial species
- Home territory ≈ 900 hectares
- Adult survival rate ≈ 0.91
- Birth rate and juvenile survival rate combined ≈ 0.20
- Juvenile dispersal process - sensitive to fragmentation
- Mean dispersal distance: 15km

Dynamic Population Model

- Model habitat as array of cells, each cell the size of a home territory
- Habitat cells are either suitable or unsuitable
 $H_{ij} = 1$ or 0
- Randomly seed adults (nesting pairs)
- Model several cycles of mortality, birth pulse, dispersal, until population reaches equilibrium
- Compute average occupancy rate for each cell as a measure of the value of the cell

Limitations of Dynamic Model

- Captures species life history dynamics
- Not useful for constrained optimization
- Need an easy to evaluate function that measures the “suitability” of a habitat for the survival of the species

Index Function Modeling Problem

- Develop an explicit landscape function (HCI) that measures habitat “values” similar to the dynamic population model (i.e., calculate occupancy rates)
- Use the explicit function to formulate constrained optimization problems in resource management

The Habitat Connectivity Index (HCI)

- The HCI is calculated as a “colonization probability”
- Each suitable cell has an independent probability of colonizing the focal cell
- The probability of colonization is a function of distance

Colonization

- Probability that a dispersing juvenile will claim a specific habitat cell
- If the location of the natal cell is (k, l) then the probability that a dispersal attempts to claim cell $(k + i, l + j)$ is p_{ij} (depends on distance d_{ij} , but independent of k, l)
- The probability that at least one neighboring cell manages to colonize cell (i, j) is given by

$$1 - \prod_{k,l} (1 - p_{kl} H_{i-k, j-l})$$

Colonization

- HCI for cell (i, j) is the product of the habitat value for that cell, H_{ij} , and the probability that at least one (other) cell will colonize (i, j) :

$$HCI_{ij} = H_{ij} \left[1 - \prod_{k,l} (1 - p_{kl} H_{i-k, j-l}) \right]$$

- Recall $H_{i-k, j-l} = 0$ or 1 ,

$$\ln(1 - p_{kl} H_{i-k, j-l}) = H_{i-k, j-l} \ln(1 - p_{kl})$$

- Let $L_{kl} = \ln(1 - p_{kl})$ (*log-dispersal propensity*)

Habitat Connectivity Index

$$HCI_{ij} = H_{ij} \left(1 - e^{-\sum_k \sum_l (H_{i-k, j-l} L_{kl})} \right)$$

- L_{kl} is obtained through the tuning process (regression)
- For the entire landscape, $HCI = \sum_{\text{all } i,j} HCI_{ij}$
- Highly nonlinear

Example Calculation

0	1	1	1	0	1
1	1	1	1	0	0
1	1	1	1	0	0
1	1	1	0	0	0
1	1	1	0	0	0
0	1	1	1	0	0

$L =$

-0.3	-0.5	-0.3
-0.5	0	-0.5
-0.3	-0.5	-0.3

$$HCI_{ij} = 1 \times (1 - e^{-0.3-0.5-0.5-0.3}) \approx 0.8$$

The Motivating Problem

- Want to determine which areas of the habitat to maintain, and which ones could be cleared or utilized for other activities (e.g., logging)
- make sure that the species will survive in the altered habitat
- Develop a constrained optimization problem

$$\begin{array}{ll} \max & HCI(x_1, x_2, \dots, x_n) \\ \text{s.t.} & c_1x_1 + c_2x_2 + \dots + c_nx_n \leq G \\ & x_j \in \{0, 1\} \text{ for } j = 1, \dots, n \end{array}$$

- e.g., can maintain as green *at most* 15 candidate squares

$$\Rightarrow c_1, \dots, c_n = 1, \quad G = 15$$

Quadratic Model

- $HCI = \sum_i \sum_j H_{ij} [1 - \exp(-\sum_k \sum_l H_{i-k,j-l} L_{kl})]$
- need to use (nicer) approximations of $HCI_{ij} = H_{ij}(1 - e^{-t_{ij}})$

$$\text{where } t_{ij} = \sum_k \sum_l (H_{i-k,j-l} L_{kl})$$

- notice that $e^t \approx 1 + t$ for small t , so define

$$QHCI_{ij} = -H_{ij} \sum_k \sum_l (H_{i-k,j-l} L_{kl})$$

- $QHCI = \sum_{i,j} QHCI_{ij}$ is positive semidefinite, so *easier* to solve the optimization problem

Example Calculation – Quadratic Model

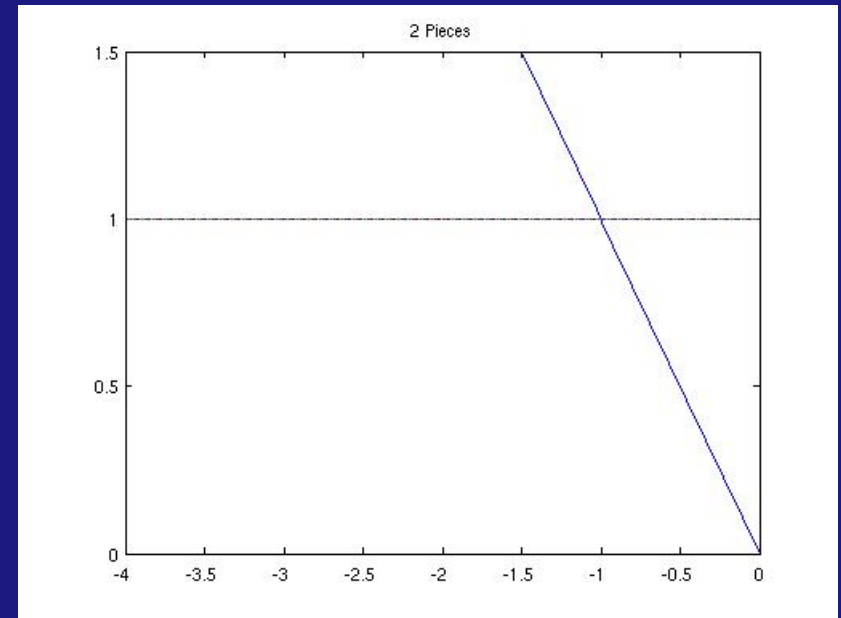
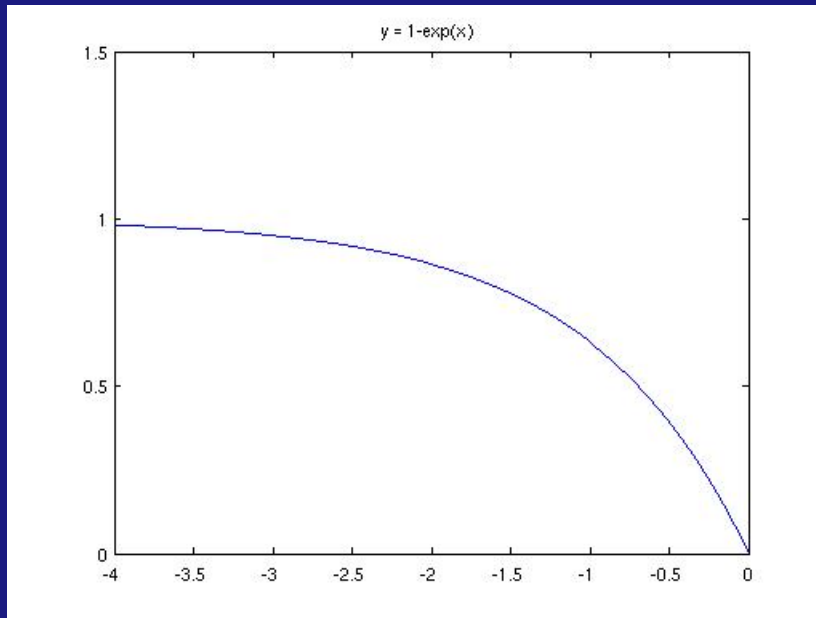
0	1	1	1	0	1
1	1	1	1	0	0
1	1	1	1	0	0
1	1	1	0	0	0
1	1	1	0	0	0
0	1	1	1	0	0

$L =$

-0.3	-0.5	-0.3
-0.5	0	-0.5
-0.3	-0.5	-0.3

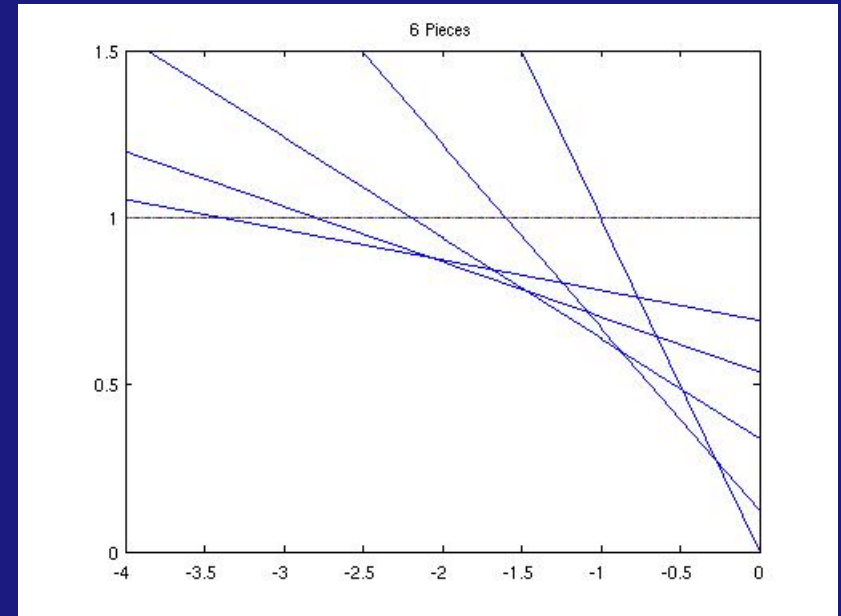
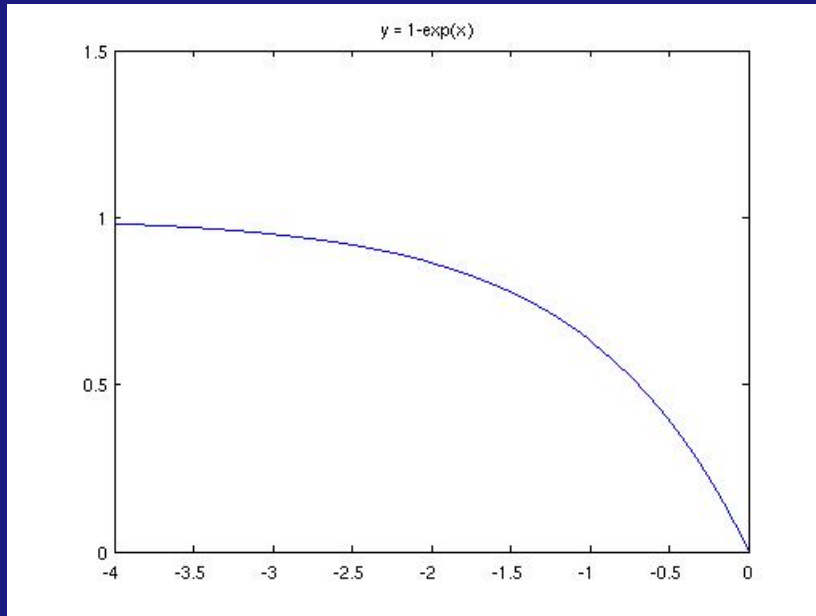
$$QHCI_{ij} = -1 \times (-0.3 - 0.5 - 0.5 - 0.3) = 1.6$$

Graph of HCI_{ij} vs $QHCI_{ij}$



- $QHCI_{ij}$ is not a good approximation of HCI_{ij}
- note that $1 - e^t \leq -t$ and also $1 - e^t \leq 1$

Linearized Model



- use *several* linear pieces to represent $1 - e^{t_{ij}}$

Linearized Model

- Let y_{ij} represent $1 - e^{-t_{ij}}$. To use t pieces, we write

$$y_{ij} \leq P_k(t_{ij}), \quad k = 1, \dots, t, \text{ where}$$
$$P_1(t_{ij}) = 1$$
$$P_2(t_{ij}) = -t_{ij}$$
$$P_3(t_{ij}) = -0.5 t_{ij} + 0.35 \text{ (say)}$$
$$\dots$$

- define $PLQHCI_{ij} = H_{ij} \times y_{ij}$, and add these constraints
- need to further linearize quadratic terms $(x_{ij} \times y_{ij}) \dots$

Fully Linearized Model

- let z_{ij} represent the quadratic term $x_{ij} \times y_{ij}$
- define $PLHCI_{ij}$ by replacing $x_{ij} \times y_{ij}$ with z_{ij} in $PLQHCI_{ij}$
- add the constraints

$$z_{ij} \leq y_{ij} + (1 - x_{ij})$$

$$z_{ij} \leq x_{ij}$$

recall: x_{ij} is 0 or 1

- $PLHCI = \sum_{i,j} PLHCI_{ij}$

Computational Results

- 125×75 landscape map, 51×51 dispersal range
- Vary # candidate habitat cells (n) and max # of candidate cells that can be kept green (G)

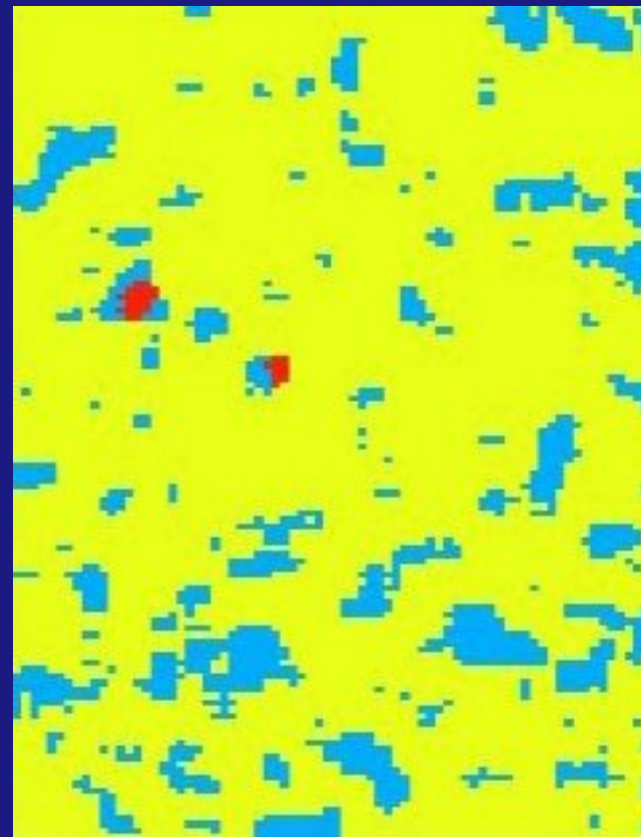
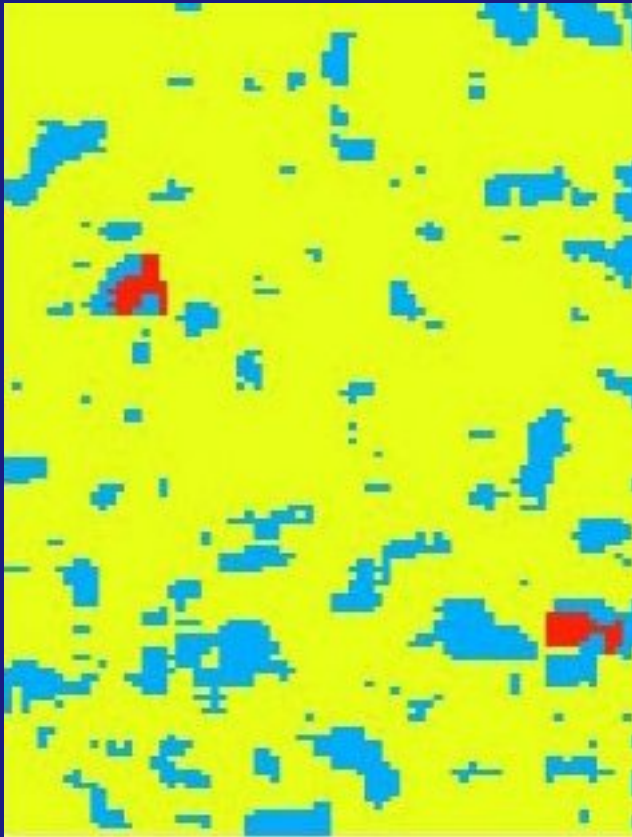
$$\begin{aligned} \max \quad & HCI(x_1, x_2, \dots, x_n) \\ \text{s.t.} \quad & x_1 + x_2 + \dots + x_n \leq G \\ & x_j \in \{0, 1\} \quad \text{for } j = 1, \dots, n \end{aligned}$$

with *QHCI* and *PLHCI* replacing *HCI*

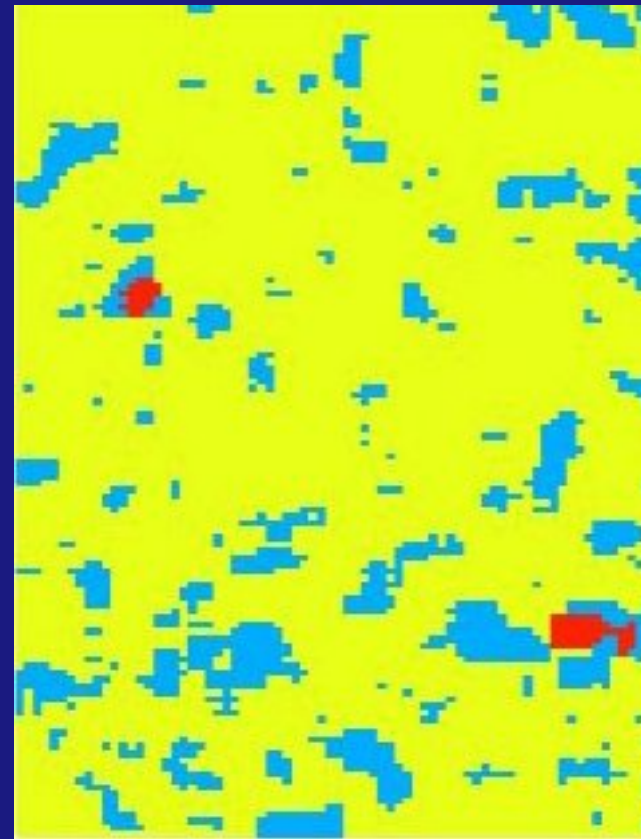
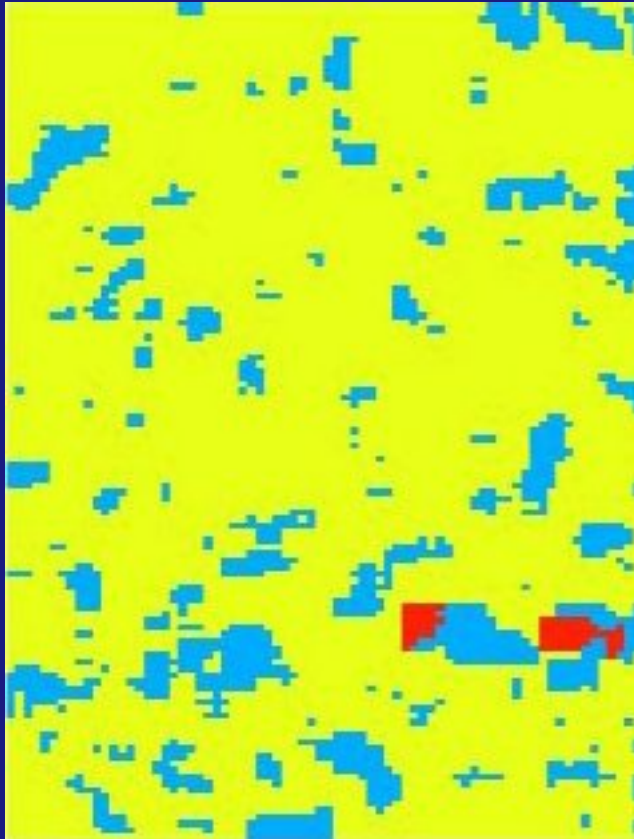
- Using AMPL-CPLEX software (state-of-the-art)
- Note running times and # branch-and-bound (BB) nodes

Candidate Habitats

- tried many different configurations of candidate habitats (red cells)



Candidate Habitats



Quadratic Model – varying n

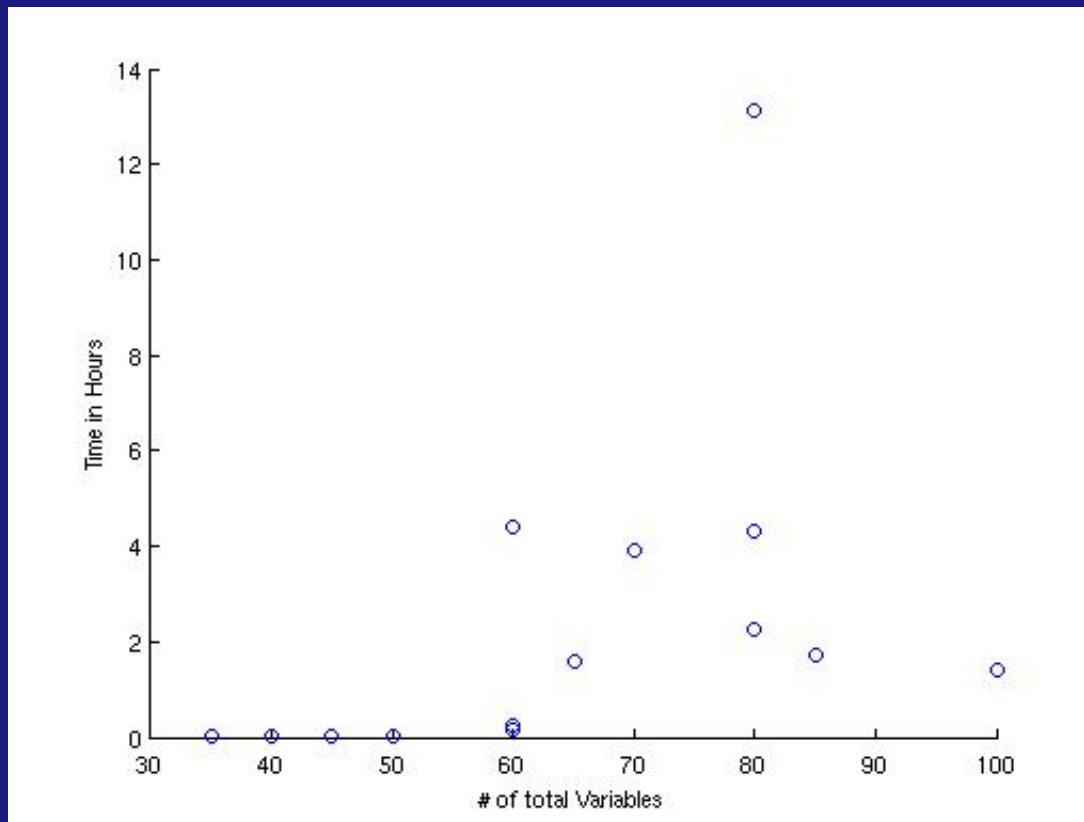


Figure 1: Run times (in hours) of the quadratic model for varying total # cells in the two (red) patches, with $G = 25$.

Quadratic Model – varying G

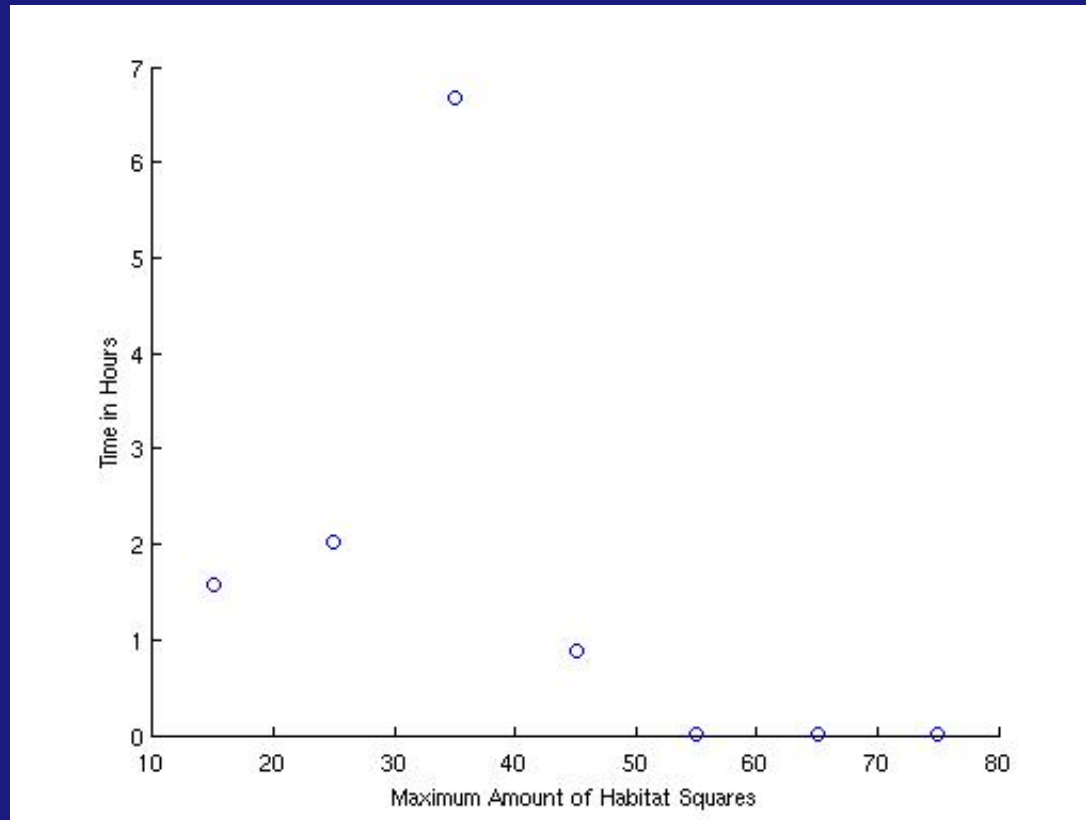
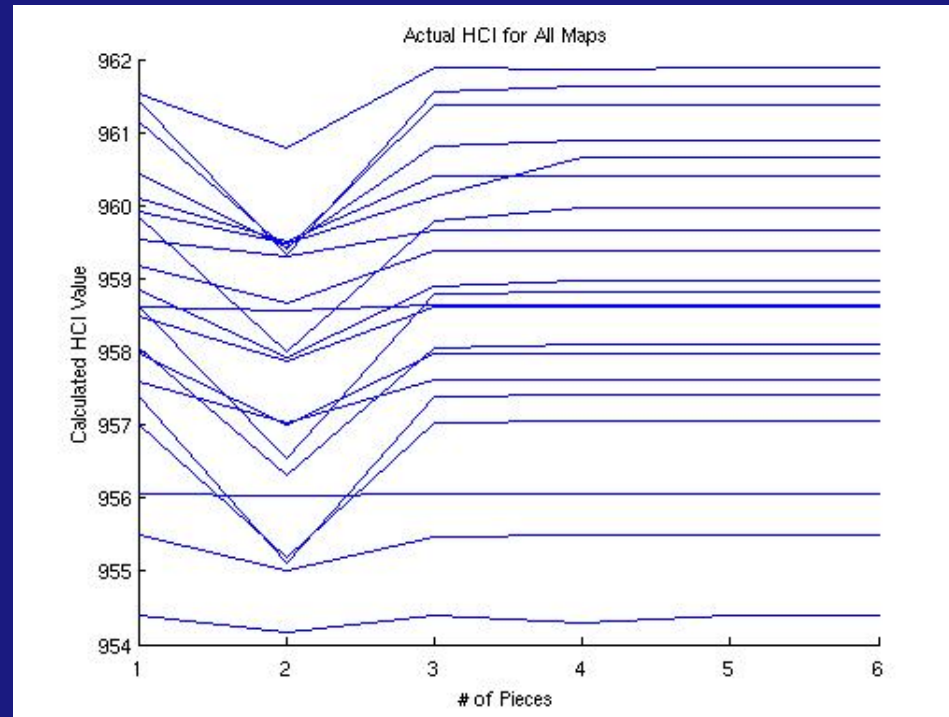


Figure 2: Run times (in hours) of the quadratic model for varying maximum # candidate cells allowed to be green, with $n = 80$.

How many pieces are needed?



- evaluate HCI at optimal solutions obtained using $PLHCI$ with various numbers of linear pieces
- 3 Pieces are sufficient

Linearized Model (3 Pieces) – varying n

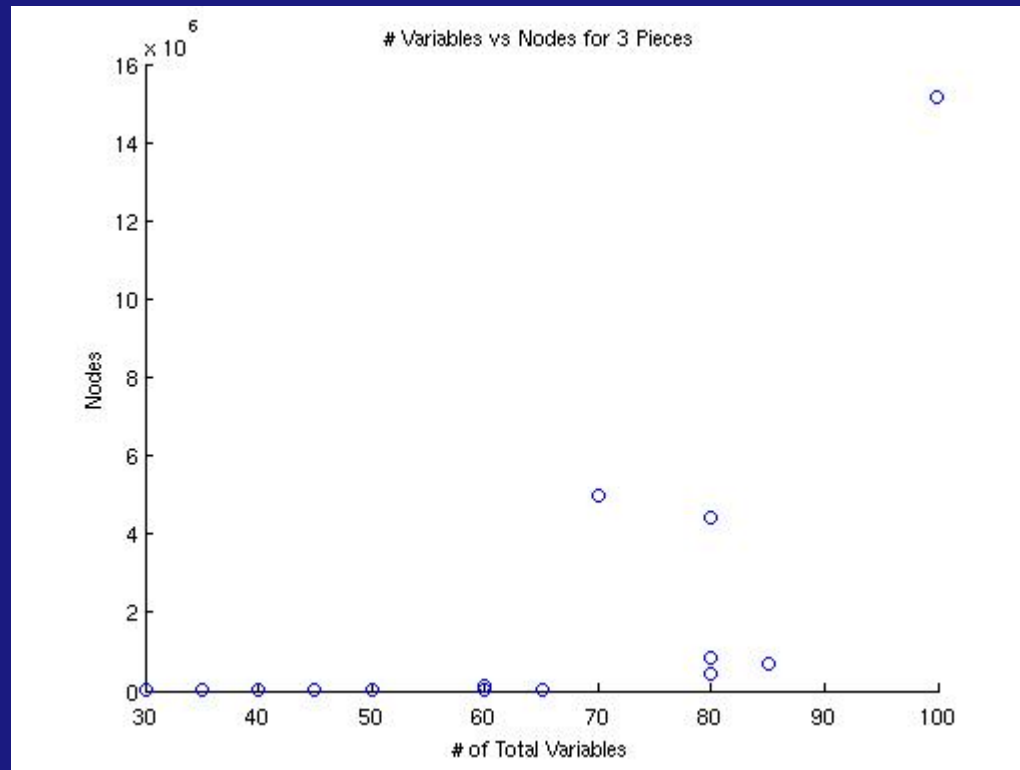


Figure 3: # BB nodes for solving the linearized model for varying total # cells in the two (red) patches, with $G = 25$.

Note: # constraints is large for large n

Linearized Model (3 Pieces) – varying G

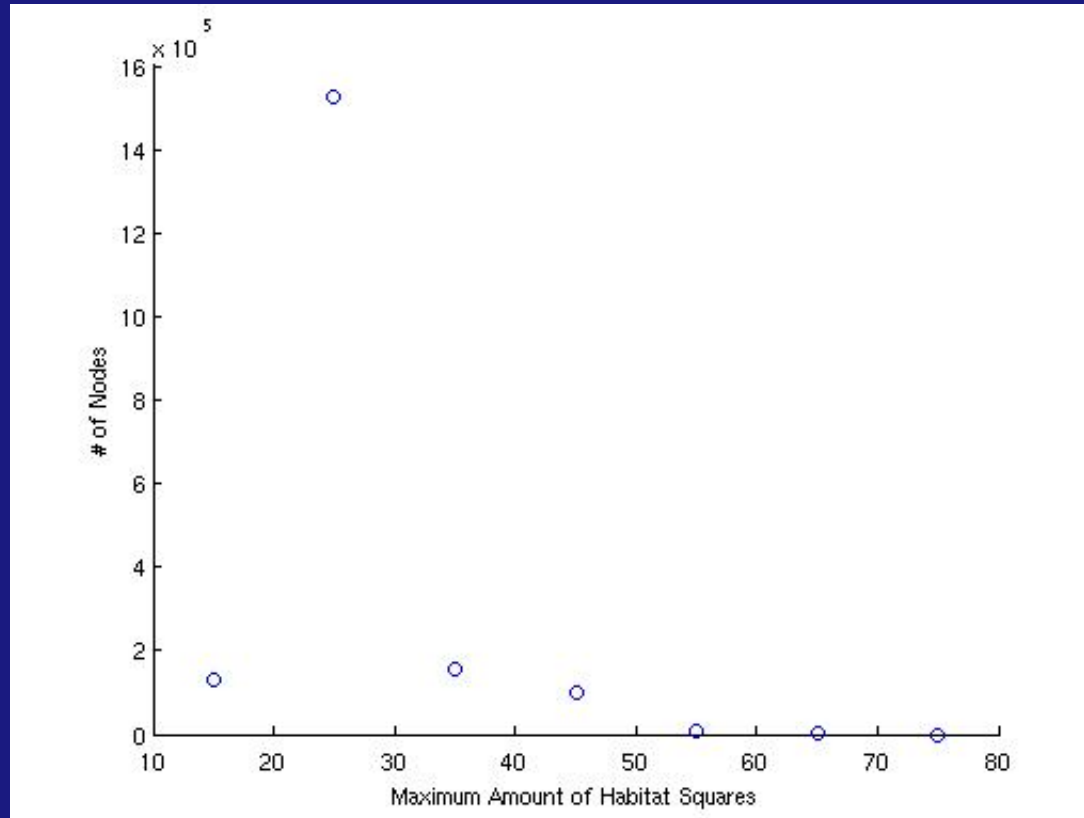


Figure 4: # BB nodes for solving the linearized model for varying maximum # candidate cells allowed to be green, with $n = 80$.

Current and Future Work

- Graded habitats – $0 \leq H_{ij} \leq 1$, with higher value indicating a more suitable cell
- Maintain HCI value (above M), while minimizing the number of candidate cells that have to be kept green:

$$\begin{aligned} \min \quad & x_1 + x_2 + \dots + x_n \\ \text{s.t.} \quad & HCI(x_1, x_2, \dots, x_n) \geq M \\ & x_j \in \{0, 1\} \quad \text{for } j = 1, \dots, n \end{aligned}$$

- Model competing species, e.g., the NSO and the barred owl