

Design and Analysis of Occupancy Studies

Part 1c

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Outline I

- 1 Single-Species Single-Season Models - Planning
- 2 Single-Species Single-Season Models - Heterogeneity
- 3 Single-Species Single-Season Models - Grand Summary

Single Species; Single-Season - Planning

Planning and Study Design

Single Species; Single-Season - Planning

Starting principles:

- Randomization - makes your sample representative of population
- Replication - controls precision
- Stratification - controls for noise, e.g. via covariates

No amount of statistical wizardry can rescue a badly executed survey.

Single Species; Single-Season - Planning

Starting principles:

- Must have some sites visited more than once to estimate detection probability
- Objectives (can) change study design:
 - Objective: Estimate overall occupancy proportion. Design: Select sites completely at random from relevant population. No control on number of sites in habitat types, so comparison of occupancy rates between habitat types may have poor power because of small number in one habitat type
 - Objective: Compare occupancy rates between habitats. Design: Stratify by habitat type, select equal number of sites from each habitat type. If you want overall occupancy, you must weight habitat occupancy by proportion of entire population of sites in each habitat type.
- Occupancy \neq Resource Selection.

Single Species; Single-Season - Planning

Defining a “site”:

- What is spatial scale were presence/absence is meaningful?
 - E.g. Remnant forest stands and rare species. Is information at stand level sufficient or do you need information on what fraction of stand is occupied?
 - E.g. Large vs. small home ranges.
- “Larger” sites have higher probabilities of occupancy than “smaller” sites (*ceteris paribus*).
 - Rule of thumb: occupancy should be $0.2 \rightarrow 0.8$ over your sites.

Single Species; Single-Season - Planning I

Site selection

- Randomize, randomize, randomize !
 - ONLY time non-random sampling acceptable is a census.
- Methods of this class assume Simple Random Sample
 - Each site has EQUAL probability of selection
 - Each site selected independently of other sites
- More complex designs possible, but beyond scope of this course (and current software)
 - Sites are forest stands of different areas and selection is proportion to size of stand (pps).
 - Sites are selected adaptively in waves, i.e. sites near where occupancy found are selected with higher probability.
 - Sites are selected using cluster and multi-state designs, e.g. random select stands, measure all trees in stand.

Single Species; Single-Season - Planning II

- AVOID selecting sites based on “prior” knowledge of occupancy UNLESS you are only interested in changes in occupancy of these sites. E.g. Selecting previously occupied stands and measuring change in occupancy over time in multi-season models.

Single Species; Single-Season - Planning

Defining a “Season”.

- Critical assumption of closure, i.e. occupancy of site does not change over season.
 - Random movement is analyzable but interpretation of “occupancy” must be modified.
 - Immigration/emigration lead to estimates of occupancy with no direct interpretation.
- Length of season depends on stability of population, i.e. slowly moving animals can be survey over longer “seasons” with closure among sites.
- “Larger” sites can have longer “Seasons” as closure more likely to be satisfied for slowly moving animals, but local deaths may be problematic.

Single Species; Single-Season - Planning I

Conducting repeat surveys.

- Many options available:
 - Visit site multiple times with a single survey per visit.
 - Visit site one with multiple INDEPENDENT surveys (e.g. different observers, different transects, different quadrats to look for fecal pellets)
- Key is that repeat surveys need to be INDEPENDENT
 - CAUTION: Detect an animals den on first visit; second visit keys on den location.
 - Use different observer on each visit who does not know location of den.
 - Use “removal” method (see later) where surveys stop after occupancy established.
 - Define “already detected” covariate in modelling

Single Species; Single-Season - Planning II

- CAUTION: Multiple simultaneous surveys with very low density (e.g. one nest per site and several transects are run) are PROBLEMATIC because if one survey detects the nest, the other survey MUST (by definition) not detect the nest.
- How will you align different surveys if models with $p(t)$ are used (e.g. think of the American Toad exercise).

Single Species; Single-Season - Planning

Conducting repeat surveys (continued).

- Avoid confounding observer/ site/ temporal effects.

		Design A			Design B		
		Day			Day		
Site		1	2	3	1	2	3
1	X X X				X		X
2				X X X		X	
3		X X X				X	X
4		X X X				X	
5			X X X			X	X
6	X X X				X		X
7		X X X			X		X
8			X X X			X	
9	X X X					X	X
p	0.5	0.3	0.8		0.5	0.3	0.8

Design B is better. [From MacKenzie et al (2006)]

- Rotate observers among sites and surveys to avoid consistent observer effects.

Single Species; Single-Season - Planning

Allocation of effort: Number of sites vs. Number of surveys.

MacKenzie and Royle (2005). J. Applied Ecology, 42, 1105-1114.
doi: 10.1111/j.1365-2664.2005.01098.x

Need to know:

- Level of acceptable precision for occupancy estimate
 - Preliminary survey: SE of 25% of $\hat{\psi}$, i.e. if $\hat{\psi} = 0.80$ then $se \approx 0.20$.
 - Management work: SE of 10% of $\hat{\psi}$, i.e. if $\hat{\psi} = 0.80$ then $se \approx 0.08$.
 - Scientific work: SE of 5% of $\hat{\psi}$, i.e. if $\hat{\psi} = 0.80$, then $se \approx 0.04$.
- Initial guess of probability of occupancy and detection. (Past surveys; other similar work).
- Resources available (e.g. what is your budget).

Single Species; Single-Season - Planning

MINIMUM number of sites needed:

Assume detectability = 100% in each site.

$$SE(\hat{\psi})_{100\% \text{detectability}} = \sqrt{\frac{\psi(1-\psi)}{s}}$$

where s is the number of sites. E.g. if $\psi \approx 0.8$ and have $s = 100$, then

$$SE(\hat{\psi})_{100\% \text{detectability}} = \sqrt{\frac{0.8(0.2)}{100}} = 0.04$$

which is acceptable for scientific work.

BUT DETECTABILITY < 100% so this is a LOWER bound on actual SE.

Single Species; Single-Season - Planning

Standard design with s surveyed sites and K surveys per site.

Total number of surveys $TS = s \times K$.

$$SE(\hat{\psi}) = \sqrt{\frac{\psi}{s} \left[(1 - \psi) + \frac{1 - p^*}{p^* - Kp(1 - p)^{K-1}} \right]}$$

where $p^* = 1 - (1 - p)^K$ is the probability of detecting species at least once in K surveys if the site is occupied.

See the spreadsheet that comes with notes.

Single Species; Single-Season - Planning

For example, what precision for $\hat{\psi}$ would be obtained when $\psi \approx 0.8$, $p \approx 0.4$, $s = 85$ and $k = 6$?

0.8	Approximate occupancy rate (psi)
0.4	Approximate detection on each SURVEY (p)
50	Number of sites (s)
6	Number of surveys/site (K)
0.95	p^* (probability of detection on a site over all K surveys)
0.0646	Estimated SE of psi-hat

Notice that $p^* = .95$ so that very few sites will be false negatives.

Single Species; Single-Season - Planning

Optimal number of surveys/site (ignoring costs) for standard design:

p	Ψ								
	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9
0.1	14	15	16	17	18	20	23	26	34
0.2	7	7	8	8	9	10	11	13	16
0.3	5	5	5	5	6	6	7	8	10
0.4	3	4	4	4	4	5	5	6	7
0.5	3	3	3	3	3	3	4	4	5
0.6	2	2	2	2	3	3	3	3	4
0.7	2	2	2	2	2	2	2	3	3
0.8	2	2	2	2	2	2	2	2	2
0.9	2	2	2	2	2	2	2	2	2

Source: MacKenzie and Royle (2005).

For very low detectability, need to take many surveys in each site!

Single Species; Single-Season - Planning

Suppose you have enough budget for 500 surveys and $\psi = 0.8$ and $p = 0.4$?

Use $K = 6$ (from table) which implies $s = 500/6 = 85$ and spreadsheet gives:

0.8	Approximate occupancy rate (psi)
0.4	Approximate detection on each SURVEY (p)
85	Number of sites (s)
6	Number of surveys/site (K)
0.95	p^* (probability of detection on a site over all K surveys)
0.0495	Estimated SE of $\hat{\psi}$

Is this precise enough?

Single Species; Single-Season - Planning

Typically costs of finding a new site (c_{new}) are $>>$ cost of each survey in a site (c_{survey}). Total cost of s sites and K surveys per site is then

$$C = s \times c_{new} + sK \times c_{survey}$$

You can use the same workbook to find the optimal choice using solver subject to constraints.

Single Species; Single-Season - Planning

Suppose $\psi \approx 0.8$, $p \approx 0.4$, $c_{new} = 10$ (hours) and $c_{survey} = 1$ (hour). The budget allowable is 2000 (hours). What is the optimal tradeoff?

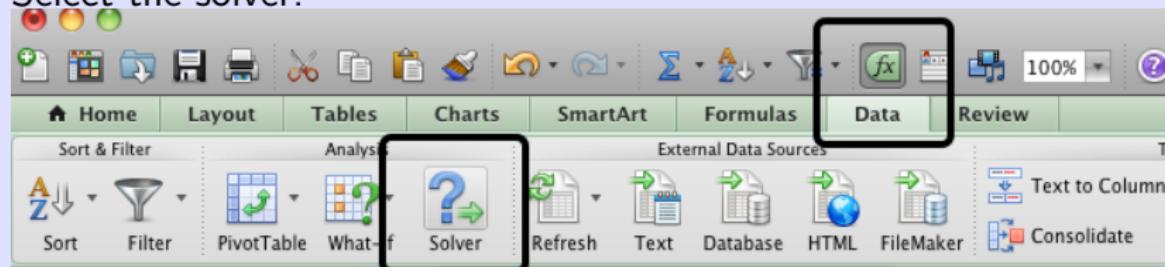
Enter basics in start of spreadsheet with a preliminary allocation of sites and surveys/site based on previous tables and allowable budget:

12	0.8	Approximate occupancy rate (psi)		
13	0.4	Approximate detection on each SURVEY (p)		
14				
15	10	Cost for a new site (c_new). Just RELATIVE COSTS are important		
16	1	Cost per survey (c_survey)		
17	2000	Maximum allowable cost		
18				
19	100	Number of sites (s)		
20	6	Number of surveys/site (K)		
21				
22	0.95	p* (probability of detection on a site over all K surveys)		
23	1600.00	Total cost		
24				
25	0.0457	Estimated SE of psi_hat		

Single Species; Single-Season - Planning

Suppose $\psi \approx 0.8$, $p \approx 0.4$, $c_{new} = 10$ (hours) and $c_{survey} = 1$ (hour). The budget allowable is 2000 (hours). What is the optimal tradeoff?

Select the solver:



Single Species; Single-Season - Planning

Suppose $\psi \approx 0.8$, $p \approx 0.4$, $c_{new} = 10$ (hours) and $c_{survey} = 1$ (hour). The budget allowable is 2000 (hours). What is the optimal tradeoff?

Complete solver dialogue box:

Solver Parameters

Set Objective: \$A\$25

To: Max Min Value Of: 0

By Changing Variable Cells: \$A\$19:\$A\$20

Subject to the Constraints: \$A\$23 <= \$A\$17

Add Change

Single Species; Single-Season - Planning

Suppose $\psi \approx 0.8$, $p \approx 0.4$, $c_{new} = 10$ (hours) and $c_{survey} = 1$ (hour). The budget allowable is 2000 (hours). What is the optimal tradeoff?

Save solutions

0.8	Approximate occupancy rate (psi)		
0.4	Approximate detection on each SURVEY (p)		
10	Cost for a new site (c_new). Just RELATIVE COSTS are important		
1	Cost per survey (c_survey)		
2000	Maximum allowable cost		
112.963208	Number of sites (s)		
7.70487963	Number of surveys/site (K)		
0.98	p* (probability of detection on a site over all K surveys)		
2000.00	Total cost		
0.0397	Estimated SE of psi-hat		

Single Species; Single-Season - Planning

Some general principles:

- Low detectability on each survey → LOTS of surveys/site!
- Low occupancy or high occupancy require fewer sites than intermediate levels of occupancy (easy to estimate 0 or 100%).
- Low occupancy implies fewer surveys/site; higher occupancy implies more surveys/site.
- Cheap surveys imply more surveys and fewer sites.
- NUMBER of SITES is the PRIMARY factor in success.

Aim to get $p(\text{detection}|\text{present})$ to be 0.80 or higher (see next few slides for details).

Single Species; Single-Season - Planning

Alternative designs: Survey some sites intensively, some sites **only once**.

Idea is that sites with many surveys estimate p and after a point, additional information is not useful and better to survey more sites.
NOT RECOMMENDED - See MacKenzie et al (2006).

- Unless $p > .8$ no benefit.
- Typically $c_{new} >> c_{survey}$ which negates taking more sites measured only once.
- May be of use when some sites are remote and dangerous to access.

Single Species; Single-Season - Planning

Alternative designs: “Removal” method.

Idea is to stop surveying sites after occupancy is (positively) confirmed. Can be more efficient than standard design

- Once species confirmed present, can you shift resources to more sites?
- Are you prepared to survey some sites longer than under standard design (uncertainty in planning)?
- Do repeat surveys use knowledge of occupancy (e.g. den) to revisit in next survey (not independent).
- CAUTION: Must assume that $p_i = p_j$ for at least one pair of (i, j) to fit a $p(t)$ model.
- Hybrid designs where some sites are surveyed completely, and some sites are surveyed until occupancy confirmed.

Single Species; Single-Season - Planning

Optimal number of (max) surveys/site for removal method.

TABLE 6.5 Optimal Maximum Number of Surveys to Conduct at Each Site for a Removal Design (K) Where All Sites Are Surveyed Until the Species Is First Detected, for Selected Values of Occupancy (ψ) and Detection Probabilities (p)

p	ψ								
	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9
0.1	23	24	25	26	28	31	34	39	49
0.2	11	11	12	13	13	15	16	19	23
0.3	7	7	7	8	8	9	10	12	14
0.4	5	5	5	6	6	6	7	8	10
0.5	4	4	4	4	4	5	5	6	8
0.6	3	3	3	3	3	4	4	5	6
0.7	2	2	2	3	3	3	3	4	5
0.8	2	2	2	2	2	2	3	3	4
0.9	2	2	2	2	2	2	2	2	3

Source: MacKenzie and Royle (2005).

Compare to previous table (slide 14).

Single Species; Single-Season - Planning

Efficiency of optimal “removal” design to standard design with equal total effort.

TABLE 6.6 Ratio of Standard Errors for Optimal Standard and Removal Designs

p	Ψ								
	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9
0.1	0.90	0.94	0.98	1.04	1.10	1.18	1.30	1.46	1.74
0.2	0.91	0.94	0.99	1.04	1.10	1.18	1.28	1.44	1.71
0.3	0.92	0.95	0.99	1.04	1.10	1.17	1.27	1.42	1.68
0.4	0.93	0.96	0.99	1.03	1.09	1.17	1.26	1.40	1.64
0.5	0.93	0.96	1.00	1.04	1.08	1.16	1.24	1.37	1.60
0.6	0.94	0.97	1.01	1.06	1.09	1.15	1.22	1.35	1.55
0.7	0.95	0.96	0.97	1.01	1.07	1.13	1.22	1.31	1.48
0.8	1.00	1.02	1.04	1.07	1.09	1.11	1.15	1.25	1.45
0.9	1.02	1.05	1.07	1.10	1.13	1.17	1.20	1.24	1.31

Values greater than 1 (in bold) indicate situations where an optimal removal design has a smaller standard error than the optimal standard design.

Source: MacKenzie and Royle (2005).

No real gain/loss unless occupancy high.

Single Species; Single-Season - Planning

General remarks:

- Increasing number sites at cost of number of surveys may not be optimal. E.g. compare $SE(\hat{\psi})$ when $\psi = 0.4$, $p = 0.3$ with $(s = 200, K = 2)$ vs. $(s = 80, K = 5)$.
- Try and reduce probability of false negative when the site is occupied to $0.05 \rightarrow 0.15$, i.e. choose K such that $0.05 < (1 - p)^K < 0.15$.
- Optimal \neq robust, i.e. violations of assumptions (e.g. heterogeneity) can cause major problems. The Standard Design is more robust than Removal Design to violations of assumptions. If heterogeneity is problem, recommend that $K \geq 3$.
- Rare species \rightarrow More sites and fewer surveys.
- To investigate hybrid designs/ robustness, simulation approach needed.
- **RUN A PILOT STUDY!**

Single Species; Single-Season - GENPRES

GENPRES: A planning tool for occupancy studies

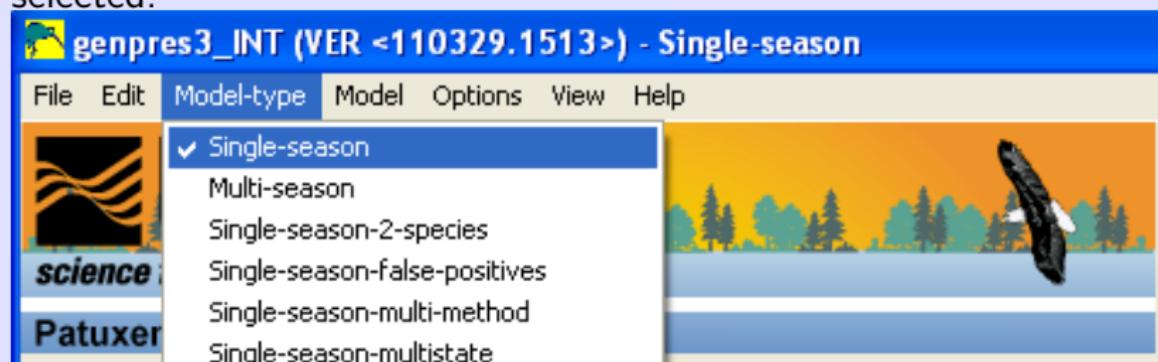
This tool allows you simulate effects of violations of assumptions, hybrid designs, or study size on the precision and bias of the estimates.

Useful for single season/multiple season; single species/ two species; etc.

Can do a bootstrap goodness-of-fit test from an analysis.
But . . . sometimes you need to combine GENPRES and PRESENCE/MARK (see multi-season models).

Single Species; Single-Season - GENPRES

Launch GENPRES; make sure Single Species; Single Site model is selected:



Single Species; Single-Season - GENPRES

What are the precision and bias if I do $s = 100$ sites and $K = 3$ with $\psi \approx 0.6$, and $p \approx 0.5$?

Enter these values in the tabbed window:

Group1

sites

100

surveys

3



PSI

.6

P(i)

.5

.5

.5

Single Species; Single-Season - GENPRES

What are the precision and bias if I do $s = 100$ sites and $K = 3$ with $\psi \approx 0.6$, and $p \approx 0.5$?

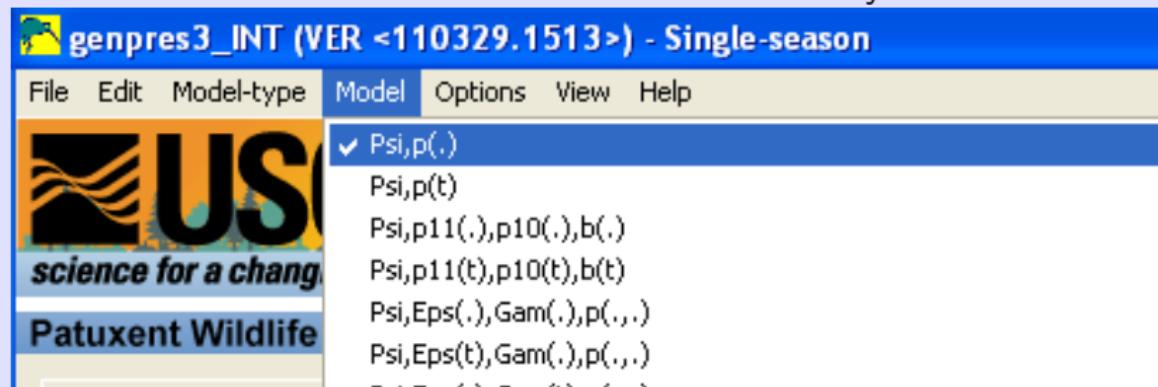
Two methods for studying design:

- Generate EXPECTED counts for each detection history.
Analyze EXPECTED counts as data. [Faster and usually sufficient, see Devineau et al (2005)], but in some cases too difficult to do (heterogeneity among animals etc).
- Generate multiple sets of simulated data, analyze each set of simulated data, look at mean/std of estimates.

Single Species; Single-Season - GENPRES

What are the precision and bias if I do $s = 100$ sites and $K = 3$ with $\psi \approx 0.6$, and $p \approx 0.5$?

Select the model with which the data will be analyzed:



Press the *Expected Value* button

Single Species; Single-Season - GENPRES

What are the precision and bias if I do $s = 100$ sites and $K = 3$ with $\psi \approx 0.6$, and $p \approx 0.5$?

```
=====
Individual site estimates of <Psi>
  site      survey     Psi    std.err   95% conf. interval
1   site_1    survey_1: 0.6000  0.0624  0.4739 - 0.7141
=====
```



```
=====
Individual site estimates of <p>
  Site      Survey     p    Std.err   95% conf. interval
1   site_1    survey_1: 0.5000  0.0493  0.4046 - 0.5954
1   site_1    survey_2: 0.5000  0.0493  0.4046 - 0.5954
1   site_1    survey_3: 0.5000  0.0493  0.4046 - 0.5954
=====
```

Estimates look to be unbiased with expected SE given. Compare to Excel Workbook results.

Single Species; Single-Season - GENPRES

What are the precision and bias if I do $s = 100$ sites and $K = 3$ but hidden heterogeneity in detectability (e.g. due to size, breeding status, coloration, etc). We now create two (or more) groups with different parameters in the approximate ratio in the population.

Group 1: $\psi \approx 0.6$, and $p \approx 0.5$ in 50 sites

Group 2: $\psi \approx 0.6$, and $p \approx 0.2$ in 50 sites

Add a group and enter the individual group parameters.

Group1	Group2
# sites 50	# surveys 3
PSI .6	PSI .6
P(i) .5 .5 .5	P(i) .2 .2 .2

Single Species; Single-Season - GENPRES

Group 1: $\psi \approx 0.6$, and $p \approx 0.5$ in 50 sites

Group 2: $\psi \approx 0.6$, and $p \approx 0.2$ in 50 sites

```
=====
Individual site estimates of <Psi>
  site      Survey      Psi   Std.err  95% conf. interval
1    site_1    survey_1: 0.5174  0.0742  0.3745 - 0.6575
=====
```

```
=====
Individual site estimates of <p>
  site      Survey      p   Std.err  95% conf. interval
1    site_1    survey_1: 0.4059  0.0584  0.2983 - 0.5233
1    site_1    survey_2: 0.4059  0.0584  0.2983 - 0.5233
1    site_1    survey_3: 0.4059  0.0584  0.2983 - 0.5233
=====
```

Are the estimates meaningful?

Single Species; Single-Season - Planning - EXERCISE I

Total budget is \$500,000. Cost per new site is \$2,000. Cost per survey is \$400.

$\psi \approx .4$, $p \approx .3$.

Plan a standard design. What is forecasted precision on $\hat{\psi}$?

Single Species; Single-Season - Planning - EXERCISE II

What is the impact of going to a hybrid design, i.e. compare

Standard design, 24 sites, 192 surveys

	Survey							
# sites	1	2	3	4	5	6	7	8
24	x	x	x	x	x	x	x	x

Panel design, 36 sites, 144 surveys

	Survey							
# sites	1	2	3	4	5	6	7	8
12	x	x	x	x	x	x	x	x
6	x	x	-	-	-	-	-	-
6	-	-	x	x	-	-	-	-
6	-	-	-	-	x	x	-	-
6	-	-	-	-	-	-	x	x

$\psi \approx .4$, $p \approx .3$. Compare results of $p(*)$ and $p(t)$ models.

Single Species; Single-Season - Planning - EXERCISE II

Standard design, 24 sites, 192 surveys

Survey								
# sites	1	2	3	4	5	6	7	8
24	x	x	x	x	x	x	x	x

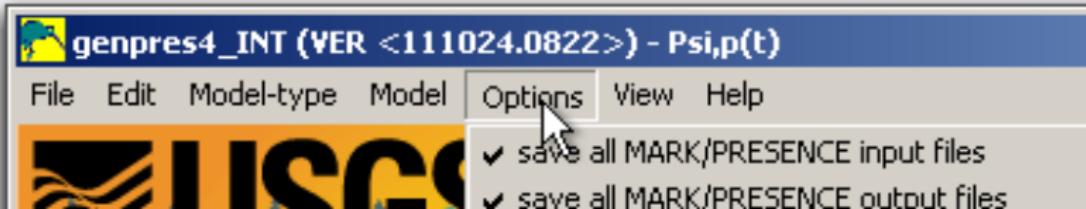
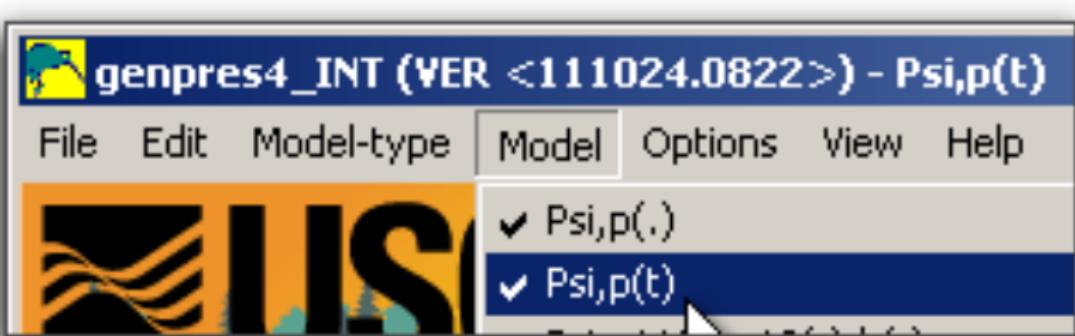
Launch GENPRES, select generating model, and complete:

The screenshot shows the GENPRES software interface with the following parameters set for 'Group1':

- # sites: 24
- # surveys: 8
- PSI: .4
- P(i): .3 (repeated 8 times)

Single Species; Single-Season - Planning - EXERCISE II

Select which model to analyze data with (you can specify more than one but it becomes messy quickly), and save all of the output (able to reproduce everything). Don't forget to save your simulation settings using File → Save.



Single Species; Single-Season - Planning - EXERCISE II

Scroll through the output labelled *mark.out* (you likely want to rename the file) and find the expected values and the “estimates”.

```
modtype-->1      1: 0 0 0 0 0 0 0 0 0 14.953421
                  2: 0 0 0 0 0 0 0 1 0.237180
                  3: 0 0 0 0 0 0 1 0 0.237180
                  4: 0 0 0 0 0 0 1 1 0.101649
                  5: 0 0 0 0 0 1 0 0 0.237180
                  6: 0 0 0 0 0 1 0 1 0.101649
                  7: 0 0 0 0 0 1 1 0 0.101649
                  8: 0 0 0 0 0 1 1 1 0.042564
```

Std Design p(*) model

Individual Site estimates of <psi>					
	Site	estimate	Std.err	95% conf. interval	
psi	1 site_1	:	0.4000	0.1063	0.2187 - 0.6136

Individual Site estimates of <p1>					
	Site	estimate	Std.err	95% conf. interval	
p1	1 site_1	:	0.3000	0.0588	0.1984 - 0.4260
p2	1 site_1	:	0.3000	0.0588	0.1984 - 0.4260
p3	1 site_1	:	0.3000	0.0588	0.1984 - 0.4260
p4	1 site_1	:	0.3000	0.0588	0.1984 - 0.4260
p5	1 site_1	:	0.3000	0.0588	0.1984 - 0.4260

Single Species; Single-Season - Planning - EXERCISE II

What is effect of going to $\psi()$, $p(t)$ model?

StdDesign p(t) model

Individual site estimates of <psi>					
	Site	estimate	Std.err	95% conf.	interval
psi	1 site_1	: 0.4000	0.1063	0.2187	- 0.6136

Individual site estimates of <p1>					
	Site	estimate	Std.err	95% conf.	interval
p1	1 site_1	: 0.3000	0.1503	0.0953	- 0.6355
p2	1 site_1	: 0.3000	0.1503	0.0953	- 0.6355
p3	1 site_1	: 0.3000	0.1503	0.0953	- 0.6355
p4	1 site_1	: 0.3000	0.1503	0.0953	- 0.6355
p5	1 site_1	: 0.3000	0.1503	0.0953	- 0.6355
p6	1 site_1	: 0.3000	0.1503	0.0953	- 0.6355
p7	1 site_1	: 0.3000	0.1503	0.0953	- 0.6355
p8	1 site_1	: 0.3000	0.1503	0.0953	- 0.6355

Single Species; Single-Season - Planning - EXERCISE II

Panel design, 36 sites, 144 surveys

# sites	Survey							
	1	2	3	4	5	6	7	8
12	x	x	x	x	x	x	x	x
6	x	x	-	-	-	-	-	-
6	-	-	x	x	-	-	-	-
6	-	-	-	-	x	x	-	-
6	-	-	-	-	-	-	x	x

Setup 5 groups with $p_s = 0$ if a site not surveyed.

Don't forget to change the number of sites visited.

Single Species; Single-Season - Planning - EXERCISE II

Group1	Group2	Group3	Group4	Group5			
# sites	12	# surveys	8	<input type="button" value="•"/>			
PSI	.4						
P(i)	.3	.3	.3	.3	.3	.3	.3

Group1	Group2	Group3	Group4	Group5			
# sites	6	# surveys	8	<input type="button" value="•"/>			
PSI	.4						
P(i)	.3	.3	0	0	0	0	0

Single Species; Single-Season - Planning - EXERCISE II

Look at expected values to check that doing what you expect:

```
modtype-->1      1: 0 0 0 0 0 0 0 0 0 7.476710
2: 0 0 0 0 0 0 1 0.118590
3: 0 0 0 0 0 0 1 0 0.118590
4: 0 0 0 0 0 0 1 1 0.050824
5: 0 0 0 0 0 1 0 0 0.118590
6: 0 0 0 0 0 1 0 1 0.050824
7: 0 0 0 0 0 1 1 0 0.050824
8: 0 0 0 0 0 1 1 1 0.050824
```

```
257: 0 0 -1 -1 -1 -1 -1 -1 -1 4.776000
258: 0 1 -1 -1 -1 -1 -1 -1 -1 0.504000
259: 1 0 -1 -1 -1 -1 -1 -1 -1 0.504000
260: 1 1 -1 -1 -1 -1 -1 -1 -1 0.216000
261: -1 -1 0 0 -1 -1 -1 -1 -1 4.776000
262: -1 -1 0 1 -1 -1 -1 -1 -1 0.504000
263: -1 -1 1 0 -1 -1 -1 -1 -1 0.504000
264: -1 -1 1 1 -1 -1 -1 -1 -1 0.216000
```

Single Species; Single-Season - Planning - EXERCISE II

What is impact of Hybrid Design with $p(*)$ model?

Hybrid Design $p(*)$ model

Individual Site estimates of <psi>					
	Site	estimate	Std.err	95% conf. interval	
psi	1 site_1	: 0.4000	0.1200	0.2002 - 0.6397	
=====					
Individual Site estimates of <p1>					
	Site	estimate	Std.err	95% conf. interval	
p1	1 site_1	: 0.3000	0.0757	0.1746 - 0.4648	
p2	1 site_1	: 0.3000	0.0757	0.1746 - 0.4648	
p3	1 site_1	: 0.3000	0.0757	0.1746 - 0.4648	
p4	1 site_1	: 0.3000	0.0757	0.1746 - 0.4648	
p5	1 site_1	: 0.3000	0.0757	0.1746 - 0.4648	
p6	1 site_1	: 0.3000	0.0757	0.1746 - 0.4648	
p7	1 site_1	: 0.3000	0.0757	0.1746 - 0.4648	
p8	1 site_1	: 0.3000	0.0757	0.1746 - 0.4648	

Single Species; Single-Season - Planning - EXERCISE II

What is impact of Hybrid Design with $p(t)$ model?

Hybrid Design p(t) model

Individual site estimates of <psi>					
	site	estimate	std.err	95% conf. interval	
psi	1 site_1	: 0.4000	0.1200	0.2002 - 0.6397	

Individual site estimates of <p1>					
	site	estimate	std.err	95% conf. interval	
p1	1 site_1	: 0.3000	0.1804	0.0737 - 0.6977	
p2	1 site_1	: 0.3000	0.1804	0.0737 - 0.6977	
p3	1 site_1	: 0.3000	0.1804	0.0737 - 0.6977	
p4	1 site_1	: 0.3000	0.1804	0.0737 - 0.6977	
p5	1 site_1	: 0.3000	0.1804	0.0737 - 0.6977	
p6	1 site_1	: 0.3000	0.1804	0.0737 - 0.6977	
p7	1 site_1	: 0.3000	0.1804	0.0737 - 0.6977	
p8	1 site_1	: 0.3000	0.1804	0.0737 - 0.6977	

Single Species; Single-Season - Planning - EXERCISE II

How many sites would you need to match expected precision from standard design?

Single Species; Single-Season - Heterogeneity

Dealing with (Hidden) Heterogeneity in
Detection Probabilities

Single Species; Single-Season - Heterogeneity

Causes of heterogeneity in detection probabilities

- Different habitats, but use habitat then as covariate.
- Different effort, but use models with p_t or covariates.
- (Hidden) different abundance in sites. Sites with higher abundance had higher chance of detecting an “occupation” than sites with lower abundance. Abundance is generally unknown (latent).
- (Hidden) differences among individuals (behaviour, size, home ranges). These cannot be measured
- (Hidden) differences among sites.

Pure Heterogeneity generally causes a negative bias in estimates of occupancy.

Single Species; Single-Season - Heterogeneity

Individual/Site (hidden) heterogeneity - approaches:

- Mixture approach – 2 or more (hidden) groups of animals/sites with different detection rates. Consequences, more histories than expected of the form 0000 or 1111.
- Distribution of detectability approach – animal/site detection rates modeled by distribution. Consequences, more histories than expected of the form 0000 or 1111.

In both cases, the group members is NOT known (latent).

Single Species; Single-Season - Heterogeneity - MARK

Finite Mixture Approach - Bull Frog Example

500 ponds were visited at 4 occasions and the observer spent 30 minutes listening for call.

Open MARK, and import the *bullfrog.inp* data in the usual fashion. Fit the $\psi(*)$, $p(*)$ model.

The screenshot shows the MARK software interface with two open windows for estimation.

The top window is titled "Detection Probability (p) Group 1 of Occupancy Estimation with Detection < 1". It contains a toolbar with icons for file operations, a table with four entries all set to 1, and a message box indicating that 5 observations are available for estimation.

1	1	1	1
---	---	---	---

The bottom window is titled "Occupancy (Psi) Group 1 of Occupancy Estimation with Detection < 1". It also has a toolbar with file icons, a table with one entry set to 5, and a message box indicating that 5 observations are available for estimation.

5

Single Species; Single-Season - Finite Mixture Models - MARK

$\psi(*)$, $p(*)$ model results

Model	AICc	Delta AICc	AICc Weight	Model Likelihood	No. Par.	Deviance
{ $\psi(*)$ $p(*)$ }	1891.2920	0.0000	1.00000	1.0000	2	33.0193

Bull Frogs

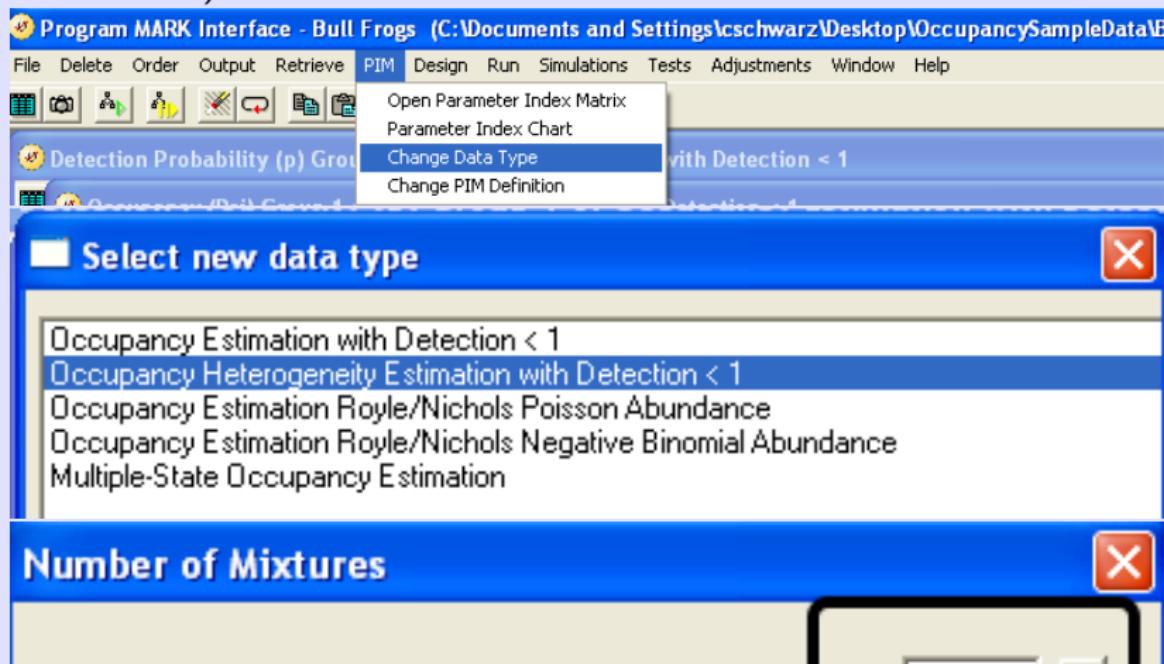
Real Function Parameters of { $\psi(*)$ $p(*)$ }

Parameter	Estimate	Standard Error	95% Confidence Interval	
			Lower	Upper
1:p	0.5519311	0.0183166	0.5158217	0.5875012
2:Psi	0.4647319	0.0233824	0.4193258	0.5107302

But the Bootstrap GOF gives evidence of a poor fit
(goodness-of-fit p-value < 0.01).

Single Species; Single-Season - Finite Mixture Models - MARK

Change the data type to a Finite-Mixture Model (usually 2 groups is sufficient):



Single Species; Single-Season - Finite Mixture Models - MARK

There is a mixing proportion (π) (why only 1 value?) and detection has two rows corresponding to the two (latent) groups.

The screenshot shows three stacked estimation tables in the MARK software:

- Prob. of Mixture (pi) Group 1 of Occupancy Heterogeneity Estimation with Detection < 1**: Shows a single value of 1 in the first row.
- Detection Probability (p) Group 1 of Occupancy Heterogeneity Estimation with Detection < 1**: Shows two rows of values: 2, 2, 2, 2 in the top row and 3, 3, 3, 3 in the bottom row.
- Occupancy (Psi) Group 1 of Occupancy Heterogeneity Estimation with Detection < 1**: Shows a single value of 10 in the first row.

Each table has a toolbar above it with various icons for model selection, estimation, and output.

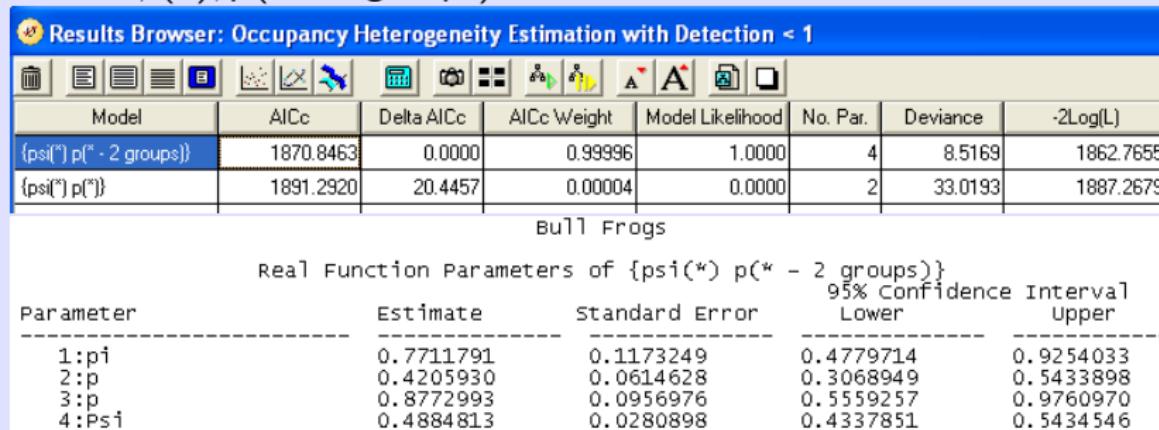
Run this model $\psi(*), p(* - 2 \text{ groups})$.

May need to try different initial values if $\hat{\psi}$ wanders off to 1 or 0.

Single Species; Single-Season - Finite Mixture Models - MARK

Model $\psi(*)$, $p(*)$ - 2 groups)

Results Browser: Occupancy Heterogeneity Estimation with Detection < 1



The screenshot shows the MARK software interface. At the top is a toolbar with various icons. Below it is a table of model selection results for 'Bull Frogs'. The table has columns for Model, AICc, Delta AICc, AICc Weight, Model Likelihood, No. Par., Deviance, and -2Log(L). Two models are listed: one with 2 groups and one with 1 group. The model with 2 groups is selected. Below the table is another table titled 'Real Function Parameters' for the selected model, showing estimates, standard errors, and 95% confidence intervals for parameters 1:pi, 2:p, 3:p, and 4:Psi.

Model	AICc	Delta AICc	AICc Weight	Model Likelihood	No. Par.	Deviance	-2Log(L)
{psi(*) p(*) - 2 groups}	1870.8463	0.0000	0.99996	1.0000	4	8.5169	1862.7655
{psi(*) p(*)}	1891.2920	20.4457	0.00004	0.0000	2	33.0193	1887.2679

Real Function Parameters of {psi(*) p(*) - 2 groups}				
Parameter	Estimate	Standard Error	95% Confidence Interval	
			Lower	Upper
1:pi	0.7711791	0.1173249	0.4779714	0.9254033
2:p	0.4205930	0.0614628	0.3068949	0.5433898
3:p	0.8772993	0.0956976	0.5559257	0.9760970
4:Psi	0.4884813	0.0280898	0.4337851	0.5434546

Deviance is much improved (unfortunately Bootstrap GOF not yet available). ΔAIC is large relative to other model.

Careful not to OVER INTERPRET the results – groups is just a DEVICE to account for heterogeneity – they often have no physical interpretation.

Single Species; Single-Season - Royle-Nichols models

Suppose that there are N_i individuals in a site, then the probability of a detection is

$$p(N_i, r) = 1 - (1 - r)^{N_i}$$

where r is the individual detection probability. Note that N_i is not observed, but if you are willing to assume that N_i comes from some distribution (e.g. Poisson with parameter λ (average density/site)), you can “integrate” over the hidden N_i .

Now $\psi = 1 - e^{-\lambda}$. Notice that ψ is a DERIVED parameter and does NOT appear in the model directly.

CAUTION: This class of models is VERY sensitive to assumptions about the distribution of the N_i .

Single Species; Single-Season - Royle-Nichols models

Change data type

The screenshot shows the PPE software interface with the following details:

- Menu Bar:** File, Delete, Order, Output, Retrieve, PIM, Design, Run, Simulations, Tests.
- Toolbar:** Includes icons for Open, Save, Help, and Results Browser.
- Results Browser:** Occupancy
- Submenu for PIM:** Open Parameter Index Matrix, Parameter Index Chart, Change Data Type (selected), Change PIM Definition.
- Dialog Box:** Select new data type
 - Occupancy Estimation with Detection < 1
 - Occupancy Heterogeneity Estimation with Detection < 1
 - Occupancy Estimation Royle/Nichols Poisson Abundance** (selected)
 - Occupancy Estimation Royle/Nichols Negative Binomial Abundance
 - Multiple-State Occupancy Estimation

Single Species; Single-Season - Royle-Nichols models

Royle-Nichols Poisson model: PIM is very simple! Notice that there is NO PIM for ψ as it is a derived parameter.

The screenshot shows a software interface for occupancy estimation. At the top, a blue header bar displays the title "Single-Species Single-Season Models - Royle-Nichols". Below the header, there are two windows stacked vertically. The top window has a blue header titled "Individual Detection Probability (r) Group 1 of Occupancy Estimation". It features a toolbar with icons for file operations (New, Open, Save, Print, Copy, Paste, Find, Delete, Undo, Redo) and a status bar at the bottom. The main area contains a table with one row and one column, labeled "1". The bottom window has a blue header titled "Mean Population Size (Lambda) Group 1 of Occupancy". It also has a toolbar with the same set of icons. Its main area contains a table with one row and one column, labeled "2". Both windows have a standard Windows-style title bar and border.

Single Species; Single-Season - Royle-Nichols models

{psi(") p(") - 2 groups}	1870.8463	0.0000	0.86395	1.0000	4	8.5169	1862.7655
(Royle-Nichols Poisson Model)	1874.5438	3.6975	0.13601	0.1574	2	16.2710	1870.5197
{psi(") p(")}	1891.2920	20.4457	0.00003	0.0000	2	33.0193	1887.2679
{psi(") p(i)}	1893.9806	23.1343	0.00001	0.0000	5	29.6104	1883.8591

Bull Frogs

Real Function Parameters of {Royle-Nichols Poisson Model}

Parameter	Estimate	Standard Error	95% Confidence Interval	
			Lower	Upper
1:r	0.4561290	0.0230632	0.4114064	0.5015705
2:Lambda	0.6591343	0.0483081	0.5592245	0.7466571

Bull Frogs

Estimates of Derived Parameters

Occupancy Estimates of {Royle-Nichols Poisson Model}

Group	Psi-hat	Standard Error	95% Confidence Interval	
			Lower	Upper
1	0.4827010	0.0249897	0.4337212	0.5316809

E(p-hat) Estimates of {Royle-Nichols Poisson Model}

Group	E(p-hat)	Standard Error	95% Confidence Interval	
			Lower	Upper
1	0.5379383	0.0211306	0.4965222	0.5793544

Single Species; Single-Season - Heterogeneity Summary

Summary about heterogeneity in catchability:

- Usually leads to underestimation of ψ .
- Largest effects when $p \approx 0$!
- Mixture-models require substantial data to work well.
- Better to design studies to minimize heterogeneity.
- Measure relevant covariates.
- If detectability related to abundance and abundance varies considerably, occupancy modeling not my first choice as estimates are (highly) sensitive to assumptions and minor changes in the data.

Single Species; Single-Season - Summary

Single-Species Single-Season Summary

Single Species; Single-Season - Summary

Planning.

- Key parameter ψ ; nuisance parameter $p_t < 1$.
- Design your study well.
 - What is appropriate spatial scale?
 - Simple Random sample of sites; some relaxation if comparing occupancy between classes.
 - What is a season – assume closure over season.
 - Repeated surveys must be independent.
- Allocate effort between sites and surveys; usually more sites and fewer surveys (but at least 2).
- Hybrid designs may be more efficient. CAUTION about “removal” design. Consider panel designs.

Single Species; Single-Season - Summary

Key assumptions.

- ① Occupancy state of sites is constant during all single-season surveys (closure).
- ② Probability of occupancy (ψ) is equal across all sites (homogeneity).
- ③ Probability of detection (p) given occupancy is equal across all sites (homogeneity).
- ④ Detection of species in each survey of a site is independent of those on other surveys
- ⑤ Detection histories at each location are independent
- ⑥ No false positives.

Some models available that deal with violations of assumptions, but these are data hungry.

Single Species; Single-Season - Summary

Analysis

- ① Maximum Likelihood & AIC & Model average
- ② Software
 - *MARK, RMark*
 - *PRESENCE, RPresence*
 - *R package unmarked*
 - *JAGS Bayesian models.*
 - *GENPRES* for planning
- ③ Carefully think of models and biological realism.
- ④ Do not data dredge.

Garbage in → garbage out.