

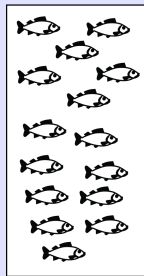
# Partial Stratification in Two-Sample Capture-Recapture Experiments

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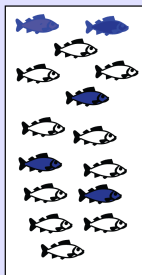
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June 24, 2014

# Basic capture-recapture



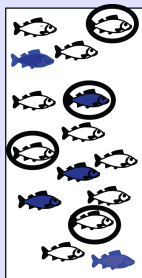
N ?



Capture  
and  
Mark  $n_1$

$n_1=100$

Fish  
Mix



Capture  
 $n_2$  of which  
 $m_2$  are marked

$n_2=200$

$m_2=10$

$$\hat{N} = \frac{n_1 n_2}{m_2} = \frac{100 \times 200}{10} = 2000$$

# Standard Lincoln-Petersen - Assumptions

- The population is **closed** (geographically and demographically).
- Mark status is correctly identified at each sampling occasion
- Marks are not lost between sampling occasions
- Capture and marking does not affect subsequent catchability of an animal
- Sampling is random
- **Homogeneous capture probabilities.**

$$\hat{N} = \frac{n_1 \times n_2}{m_2} \approx \frac{Np_1 \times Np_2}{Np_1p_2} \approx N$$

# Standard Lincoln-Petersen - Need for stratification

Animals may vary in catchability, e.g. by sex.

Sex	$N$	$p_1$	$p_2$	$n_1$	$n_2$	$m_2$	$\hat{N}$
M	1200	.10	.05	120	60	6	
F	800	.05	.10	40	80	4	
Total	2000			160	140	10	2240

Bias can go in either direction depending on correlation of catchabilities.

# Standard Lincoln-Petersen - Need for stratification

Animals may vary in catchability, e.g. by sex.

**Stratify and compute separate estimates for each stratum.**

Sex	$N$	$p_1$	$p_2$	$n_1$	$n_2$	$m_2$	$\hat{N}$
M	1200	.10	.05	120	60	6	1200
F	800	.05	.10	40	80	4	800
Total	2000						2000

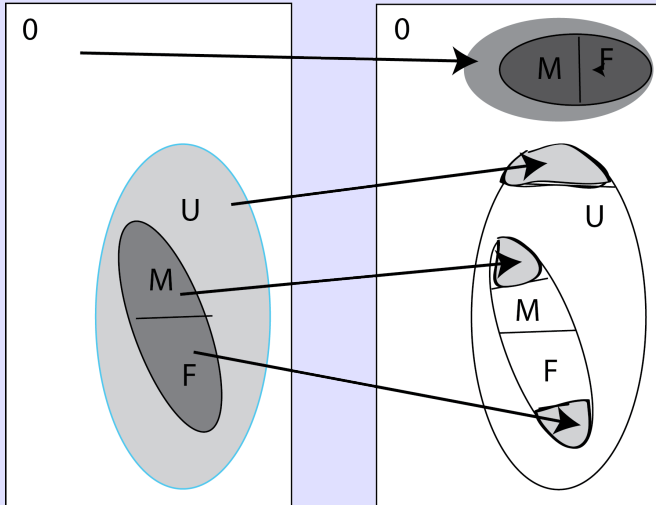
**Assumes that you can classify animals into strata in ALL sampling occasions.**

# Partial Stratification

Sometimes difficult/costly to stratify when captured.  
Select a sub-sample of each sample to stratify.

Time 1

Time 2



# Partial Stratification

$3C + 4$  possible capture histories.

History	Probability
MM	$\lambda_M p_{1M} \theta_1 p_{2M}$
M0	$\lambda_M p_{1M} \theta_1 (1 - p_{2M})$
0M	$\lambda_M (1 - p_{1M}) p_{2M} \theta_2$
...	Similarly for other categories
UU	$\lambda_M p_{1M} (1 - \theta_1) p_{2M} + \lambda_F p_{1F} (1 - \theta_1) p_{2F}$
U0	$\lambda_M p_{1M} (1 - \theta_1) (1 - p_{2M}) + \lambda_F p_{1F} (1 - \theta_1) (1 - p_{2F})$
0U	$\lambda_M (1 - p_{1M}) p_{2M} (1 - \theta_2) + \lambda_F (1 - p_{1F}) p_{2F} (1 - \theta_2)$
00	Everything else (not observable)

Where

- $p_{tC}$  = probability of capture of category  $C$  at time  $t$ .
- $\lambda_C$  = proportion of category  $C$  in the population.
- $\theta_t$  = proportion of sample at time  $t$  that is “categorized”.

Multinomial distribution with unknown index

$$L = \frac{N!}{n_{U0}! \ n_{UU}! \ n_{0U}! \dots (N-n)!} \times (P_{U0})^{n_{U0}} (P_{UU})^{n_{UU}} (P_{0U})^{n_{0U}} \times \prod_C (P_{C0})^{n_{C0}} \times \prod_C (P_{CC})^{n_{CC}} \times \prod_C (P_{0C})^{n_{0C}} \times (P_{00})^{N-n}$$



# Partial Stratification - Estimation

No closed form solution - standard numerical methods used.

Parameters can be constrained by using the design matrix and offset vectors

$$\text{logit}(p_{tk}) = X\beta + \text{offset}$$

Example, a model with  $p_{1M} = p_{1F}$  ,  $p_{2M} = p_{2F}$  uses design matrices and offsets

$$\begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \\ 0 & 1 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Use Akaike's information criterion ( $AIC$ ) ( Burnham and Anderson, 2002) to rank models.

# Partial Stratification - Optimal Allocation

Use linear cost function

$$C = n_1 c_1 + n_1^* c_1^* + n_2 c_2 + n_2^* c_2^* \leq C_0$$

where

- $n_1$  is number of fish captured at first occasion;
- $n_1^*$  is number of fish captured at first occasion that are “categorized”.  $n_1 \geq n_1^*$ .
- $n_2$  is the number of fish captured at second occasion;
- $n_2^*$  is the number of fish captured at second occasion that are “categorized”.  $n_2 \geq n_2^*$ .

# Partial Stratification - Example

## Walleye Data-Mille Lacs, MN

- Walleye are captured on the spawning grounds. Almost all the fish can be sexed in the first occasion
- All the captured fish are tagged and released and the recapture occurred 3 weeks later using gill-nets
- From a sample of fish captured at second occasion that are not tagged, a random sample is selected and sexed

Capture History	statistics
<i>U0</i>	40
<i>UU</i>	1
<i>M0</i>	5067
<i>MM</i>	40
<i>F0</i>	1551
<i>FF</i>	33
<i>0M</i>	41
<i>0F</i>	237
<i>0U</i>	3075

# Partial Stratification - Example

Model			np	$\hat{N}$ '000s	s.e.( $\hat{N}$ ) '000s	$\Delta AIC_c$
$p(C * t)$	$\theta(t)$	$\lambda(C)$	8	205	26	0.0
$p(C * t)$	$\theta(t)$	$\lambda(.)$	7	208	24	4.7
$p(t)$	$\theta(t)$	$\lambda(C)$	6	311	36	462.2
$p(C)$	$\theta(t)$	$\lambda(C)$	6	399	54	1561.5
$p(.)$	$\theta(t)$	$\lambda(C)$	5	348	39	1572.0
$p(C)$	$\theta(t)$	$\lambda(.)$	5	399	54	11613.4
$p(.)$	$\theta(.)$	$\lambda(.)$	3	348	39	13276.8

- $p$  = probability of capture
- $\lambda$  = proportion of category in the population.
- $\theta$  = proportion of sample at time that is "categorized".
- $C * t$  = varies by category and time;
- $t$  = varies by time but not category;
- $C$  = varies by category but not by time
- $.$  = no variation by time or by category.

# Partial Stratification - Example

Parameter	MLE	SE	SE if sex ratio known
$\lambda_M$	.33	.06	-
$\lambda_F$	.67	.06	-
$\theta_1$	.993	.001	.001
$\theta_2$	.082	.004	.004
$N$	205,000	26,000	25,000
$N_M$	68,000	13,000	8,000
$N_F$	137,000	23,000	17,000
$\hat{N}_{LP}$	311,000	36,000	

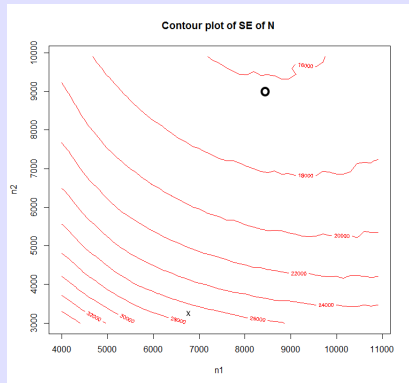
## Example - optimal allocation

Let  $c_1 = 2$ ,  $c_2 = c_1/2$ ,  $c_1^* = 4$ ,  $c_2^* = c_1^*/2 = 4$ ,  $C_0 = 40000$

$n_1 = 8199$   $n_1^* = 3003$   $n_2 = 9511$   $n_2^* = 997$

$SE(\hat{N}) = 16000$  a 30% reduction.

Conditional Contour plot and Graph for  $SE(\hat{N})$ :



# Partial Stratification - Summary

## Problem

- Capture heterogeneity can cause bias in estimates in capture-recapture experiments
- Stratification may not be possible/is costly for all captured animals in each occasion

## Solution

- Partial stratification
- Optimal allocation

## Future work

- Bayesian solution for prior information on sex ratio
- Individual covariates such as length