#### **Methods of abundance estimation**

Why count? •Concern about conservation •Lion / tiger

•Assessment of resource base •Timber/ NTFP

•Assessment of threat to humans •Blackbuck count- farmers •Elephant count-villagers

Estimating population and its growth
Estimate age – sex composition
Compare population with carrying capacity

# Standard methods Capture- recapture Nearest neighbor

•Line transect



#### •Non- standard methods

- •Waterhole census
- •Dung pile count
- •Bird count based on calls
- Pugmarks method

# Estimation of Tiger Population using Pugmarks

 Pugmark - a single tiger paw print as a tracing / photograph taken from fixed height

•Pugmark Track Set – a series of paw prints from the same individual

#### Traditional method

- •Locate pugmarks
- •Trace them onto paper
- •Record time and place
- •Visual comparison
- •Eliminate repeats
- •What is left is the count

- Criticism
  - •No validation
  - •No oscillation in count
  - •Very low count of cubs
  - •Subjective
  - •Identifiability questionable

### How to improve? Quantify shape and size

#### • Pad

- 1. Area
- 2. Major axis
- 3. Minor axis
- Pug

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- 4. Length
- 5. Width

- Distance
  - 6. Toe 1 centre to Toe 2 centre
  - 7. Toe 2 centre to Toe 3 centre
  - 8. Toe 3 centre to Toe 4 centre
  - 9. Pad centre to Toe 1 centre
  - 10. Pad centre to Toe 2 centre
  - 11. Pad centre to Toe 3 centre
  - 12. Pad centre to Toe 4 centre

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#### A.P.Gore

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S.A.Paranjpe

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#### How to use measurements?

Establish objective criteria

#### Factors possibly affecting a pugmark print

- Substrate
- •Operator
- •Sex of animal
- •Locality
- Inter-individual variation
- Intra-individual variation

#### **Experiments**

•Substrate- same animal different substrates : Coimbatore

- •3 substrates- fine soil, wet mud, sand
- •Two animals only
- ANOVA for each variable
- Most variables show no effect

•Operators- same trail different operators: Melghat

- •Single trail
- •6 operators
- •3 prints each
- ANOVA for each variable
- Most varibles show no effect

•Sex- Pune, Coimbatore

•5 animals- 2 males, 3 females

•Several prints on each (total 100)

Logistic regression : P= prob( a given animal is a female)

•3 measurements significant

•Pad center- Toe 1 center distance(X<sub>1</sub>)

•Pad center- Toe 2 center distance (X<sub>2</sub>)

•Toe 1 center- Toe 4 center distance (X<sub>3</sub>)

•Ln(p/(1-p)) =  $56.06 - 2.76X_1 - 3.56X_2 - 1.1X_3$ 

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#### Sex identification using logistic regression

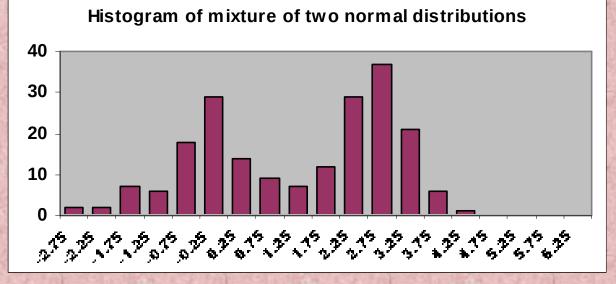
Decision	Know	Total	
	Μ	F	
M(p<0.25)	23	7	30
F(p>0.75)	1	59	60
Ambiguous	6	1	7
Total	30	67	97



#### Sex discrimination using Bhattacharya method

- Mixture of normal distributions : to be separated
- •For a single normal distribution
- Y = ln (f(x+ $\Delta$ )/ f(x)) = [ 2  $\Delta$ (x+  $\Delta$ /2) 2  $\Delta$   $\mu$ ] / 2  $\sigma^2$
- Slope of a line =  $\Delta/\sigma^2$  intercept =  $\Delta \mu / \sigma^2$
- Mixture of distributions gives multiple falling lines
- •Plot y Vs x+  $\Delta/2$ . Gives sequence of straight lines with negative slope- each line one component

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# Bhattacharya plot

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Bhattacharya method:
Male – squarish print
Female – rectangular
(Length – breadth) (of the covering frame)
Near 0 – male , > 0 female
Histogram of Breadth bimodal
Separate components by Bhattacharya method
Component with smaller mean breadth- females
With larger mean breadth – males

**Results**:

 $\mu_1 = 10.26$ ,  $\mu_2 = 12.75$ 

estimated proportion of females=  $40/75 \approx 0.53$ 

Simplified rule adopted by foresters: (L-B) >2.5cm - female

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•Locality - field data from project tiger: Melghat and Kanha

•90 prints each

Half data used for calibration

Discriminant analysis

Half data used for validation

Initially 10 variables used

Variables tested for redundancy

•Two variables adequate
•Toe1C- Toe2C distance
•Toe1C- Toe3C distance

# Classification of 90 tiger pugmark tracings from two populations

True	Classification Using					
Population	10 var.	iables	2 variables			
	I	Π	Ι	Π		
Ι	44	1	40	5		
II	8	37	4	41		

#### Does this help in estimating Tiger number?

•No.

- •All analysis so far macro level only
- Key question How many distinct tigers in a set of n pugmark tracings?
- •An intuitive approach:
  - •Compare known intra individual variation with observed inter pugmark variation.
  - How to assess intra individual variation?
  - •Analyze multiple tracings from the same trail.

#### An intuitive algorithm

- • $X_1$ ,  $X_2$ ,...,  $X_n$ : n vectors ( n pugmarks) of order p
- •<u>X</u><sub>i</sub> Follows normal distribution
   with mean vector <u>μ</u><sub>i</sub>
  •Var- cov matrix\_Σ (assumed to be known)
- •Number of distinct tigers k (unknown)
- •Step 1  $H_0$  : k = 1 ( only one tiger)
- • $T_n = sum[(X_i Xbar)' \Sigma^{-1} (X_i Xbar)]$

•Reject H<sub>0</sub> if T large .Then try k=2

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#### An intuitive algorithm(cont.)

- $H_0: k=2$
- Omit one vector  $\underline{X}_i$  which causes max reduction in  $T_n$
- Check if  $T_{n-1}(i) = sum[(X_i Xbar)' \Sigma^{-1} (X_i Xbar)]$  is small
- If yes, two tigers. Otherwise continue omitting.
- Process terminates when homogeneous subgroups identified
- Each subgroup one tiger
- confidence statement?
- •Not available in standard literature.

•Recent algorithm developed by Chatterjee and Samanta(1999) A.P.Gore •Not tested 19 S.A.Paranjpe

### Foresters' current approach Discriminant Analysis

- Establish the parameters fit for consideration to facilitate individual identity
- Multiple Group Discriminant Analysis done using SPSS Systat (unitwise)
- Canonical Scores Plot from data generated by tracings/ digital photographs in SPSS Sigma ScanPro
- Number of distinct pugmarks ascertained unitwise
- If all tracks from each foot are grouped into 2 groups (eg. A & B) : then these sets are from 2 different tigers
- Equal mixing (overlap) may indicate a single animal
- 75% to 80% correct grouping is acceptable

#### **Camera trap method**

#### Major criticism about pugmark method

#### Identifiability

- Alternative suggested : Camera trap
- Tiger walks on footpaths

#### Two cameras placed on two sides of road opposite to each other

- Activated as a laser beam is cut by any object
- Tiger picture from both sides

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- Claim : stripes on the back of tiger carry signature
- Treat photograph as capturing and marking a tiger
- •Use capture-recapture model to estimate number

Objections:
Identifiability
Cost effetiveness
Feasibility
Proper sampling of forest

•Controversy: which method is better •Camera trap or pugmark?

## Comparison Camera trap Vs Pugmark

#### •**Identity** Based on picture Ocular comparison

based on pugmark numerical comparison

#### Intra individual variability

Not measurable

measurable

•Cost Expensive Equipment intensive

#### •Suitability

Not suited for low density Unsuitable for rugged terrain inexpensive labor intensive

#### low density –no limitation unsuitable for swamps

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## Comparison Camera trap Vs Pugmark(cont.)

#### Sampling

Will see only a small fraction

virtually every individual seen. Can build a directory. Individual life history can b accumulated. no problem

Territoriality is a problem

**Confidence interval** 

Can be given

cannot be given