

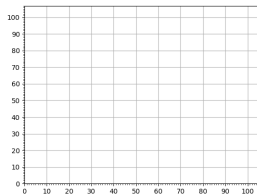
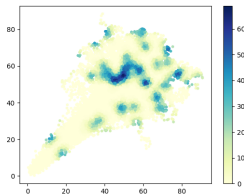
VE414 Presentation

Group 9

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Overview of Density Estimation

- Known: Many records of the number of Teyes
- Required: Estimation of Teyes in the Forest
- Solution:
 - Divide map into grid (107×107)
 - Calculate the density on the grid according to the records within 1 meter
 - For two records using *density estimation of two intersected circles*
 - For more than two records taking the mean of the estimation
 - Teyes sample according to the density

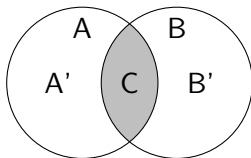


Density Estimation of two Intersected Circles

Suppose there are two detected circles which intersect with each other.
We only know

$$\text{Tayes}[A = A' \cup C] = a, \quad \text{Tayes}[B = B' \cup C] = b$$

from the magic of Hermione.



If Tayes are (approx) uniformly distributed in a circle, and $S_A \approx S_B$, $a \approx b$,

$$\text{Density}[C] \approx \left(\frac{a}{S_A} + \frac{b}{S_B} \right) / 2, \quad c = \text{Density}[C] \cdot S_C,$$

$$\text{Density}[A'] \approx \frac{a - c/2}{S_A - S_C}, \quad \text{Density}[B'] \approx \frac{b - c/2}{S_B - S_C}.$$

However, these conditions don't always hold, so we need to extend our model. Keep the assumption that Tayes are uniformly distributed in a circle, the number of Tayes x in area C should follow binomial distributions $B(a, p_a = S_C/S_A)$ and $B(b, p_b = S_C/S_B)$ in two circles.

$$P[X = x] = \binom{a}{x} p_a^x (1 - p_a)^{a-x} \binom{b}{x} p_b^x (1 - p_b)^{b-x},$$

$$P_x = P[X = x] \bigg/ \sum_{i=0}^{\min\{a,b\}} P[X = i],$$

$$\text{Density}[A'] = \frac{1}{S'_A} \sum_{i=0}^{\min\{a,b\}} (a - i) P_i, \quad \text{Density}[B'] = \frac{1}{S'_B} \sum_{i=0}^{\min\{a,b\}} (b - i) P_i,$$

$$\text{Density}[C] = \frac{1}{S_C} \sum_{i=0}^{\min\{a,b\}} i P_i.$$

Guassian Mixture Distribution

- Assumption: The distribution of the fruits from a tree follows the *Bivariate Gaussian Distribution*.
- The distribution of Tayses on the ground \sim Guassian Mixture Distribution

$$p(x) = \sum_{k=1}^K \pi_k N(x|\mu_k, \Sigma_k)$$

where $\sum_{k=1}^K \pi_k = 1$. K is the number of trees, μ_k is the position of the tree, π_k and Σ_k determines some unique properties of a tree.

- Aim: Fit Guassian Mixture Model(GMM) with Tayses Samples

Fit GMM with Expectation-Maximization algorithm (EM)

- E-step: calculate the posterior probabilities for $k \in \{1, \dots, K\}, n \in \{1, \dots, N\}$

$$\gamma_{nk} = p(k|x_n) = \frac{\pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k)}{\sum_{j=1}^K \pi_j \mathcal{N}(x_n | \mu_j, \Sigma_j)}$$

- M-step: calculate the value for π_k, μ_k, Σ_k

$$\begin{aligned}\mu'_k &= \frac{\sum_{n=1}^N \gamma_{nk} x_n}{\sum_{n=1}^N \gamma_{nk}} \\ \Sigma'_k &= \frac{\sum_{n=1}^N \gamma_{nk} (x_n - \mu'_k)(x_n - \mu'_k)^T}{\sum_{n=1}^N \gamma_{nk}} \\ \pi'_k &= \frac{\sum_{n=1}^N \gamma_{nk}}{N}\end{aligned}$$

- Loop until log likelihood convergence:

$$\ln p(X|\mu, \Sigma, \pi) = \sum_{n=1}^N \ln \left\{ \sum_{k=1}^K \pi_k \mathcal{N}(x_n | \mu_k, \Sigma_k) \right\}$$

- Pick a $k = \arg_k \min \text{BIC}(\text{Bayesian Information Criterion})$:

$$BIC = -2\ln(L) + \ln(N) * K$$

Generalize tree distribution to unobserved areas

- Assumptions

- Assumption 1: The number of trees in each grid of a certain size follows Poisson distribution with parameter λ .
- Assumption 2: In each grid, the trees are located uniformly.

- Estimate λ

$$L(\lambda) = \ln \prod f(k_i | \lambda) = -n\lambda - \sum \ln(k_i!) + (\sum k_i) \ln \lambda$$

$$\hat{\lambda}_{MLE} = \frac{1}{n} \sum k_i$$

- Random Generation

- 1 For each unobserved grid, sample the number of trees in it according to the estimated Poisson distribution
- 2 For grids with trees, randomly generate positions for these trees

Result

- Total estimated number of trees in the forest: $25 + 79 = 104$
- An example of possible tree positions

