

# Diffusion Approximation

## Effective population size estimation

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November 29, 2009

# Outline

- 1 Introduction
  - Motivation
  - What is Effective population size
- 2 Question
  - Objective
  - Issue
- 3 Solution
  - Diffusion approximation
  - Simplex expansion

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# Motivation

- Successfulness of a species

Population size is a sensible indicator

- But..... Imagine a species

6 Billion

1 Male (*Distorted Sex ratio*)

Clones (*limited genetic variation*)

- Is this so different from a species with only one female and one male left?

6 Billion  $\approx$  2

- Something better is needed

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$N_e$ 

- Effective population size  $N_e$ 
  - Abstract and conceptual population size
  - Sex ratio, Genetic variation, Breeding structure
- Ecology, Population Genetics, Conservation

# Estimation of $N_e$

- How to estimate  $N_e$  given a sample of genetic data?

- $\mathbb{E}[r^2] = f(N_e)$



$$r = \text{cor}\left( \begin{array}{cc} \# \text{Allele A}, & \# \text{Allele B} \\ \uparrow & \uparrow \\ \text{locus 1} & \text{locus 2} \end{array} \right)$$

- Sample mean  $\hat{r}^2$  can be obtained from data
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- If we know.....then .....

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- $r^2$  can be defined in terms of Genotype frequency
- Analytic derivation of the expectation by integration is not going to be easy

$$= \left( -4 \left( p_1 + p_2 + \frac{1}{2} (p_3 + p_4 + p_6 + p_7) + p_5 \right) \right. \\ \left. \left( p_1 + \frac{1}{2} (p_2 + p_4 + p_6 + p_9) + p_3 + p_8 \right) + \right. \\ \left. 4 p_1 + 2 p_2 + 2 p_3 + p_4 + p_6 \right)^2 / \\ \left( 4 \left( - \left( p_1 + p_2 + \frac{1}{2} (p_3 + p_4 + p_6 + p_7) + p_5 \right) \right)^2 - \right. \\ \left. \frac{1}{4} (2 p_1 + 2 p_2 + p_3 + p_4 + 2 p_5 + p_6 + p_7 - 2) \right. \\ \left. (2 p_1 + 2 p_2 + p_3 + p_4 + 2 p_5 + p_6 + p_7) + p_1 + p_2 + p_5 \right) \\ \left( - \left( p_1 + \frac{1}{2} (p_2 + p_4 + p_6 + p_9) + p_3 + p_8 \right) \right)^2 - \\ \left. \frac{1}{4} (2 p_1 + p_2 + 2 p_3 + p_4 + p_6 + 2 p_8 + p_9 - 2) \right. \\ \left. (2 p_1 + p_2 + 2 p_3 + p_4 + p_6 + 2 p_8 + p_9) + p_1 + p_3 + p_8 \right)$$

Figure: Rsq



# Diffusion

- Diffusion Approximation

- Diffusion Model built upon assumptions from a very popular reproduction model in population genetics, namely Wright-Fisher's Model

- Differential Operator

$$\begin{array}{ccc} \text{Diff}(G(p_1, p_2, \dots, p_9)) = & g(p_1, p_2, \dots, p_9) \\ \uparrow & \uparrow \\ \text{Any function} & \text{Special Derivative} \end{array}$$



$$\mathbb{E}[g] = 0$$

- A powerful “machine” produce all kinds of expressions of  $\mathbb{E}$  given a “recipe”  $G(p_1, p_2, \dots, p_9)$

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## Too hard

- Theoretically a careful choice of  $G(p_1, p_2, \dots, p_9)$

- But.....

$$F(\mathbb{E}[r^2]) = 0$$

$$\begin{aligned} \text{Diff} &= \frac{1}{2}x(1-x)\frac{\partial^2}{\partial x^2} + \frac{1}{2}y(1-y)\frac{\partial^2}{\partial y^2} \\ &+ \frac{1}{2}[x(1-x)y(1-y) + D(1-2x)(1-2y) - D^2]\frac{\partial^2}{\partial D^2} \\ &+ D\frac{\partial^2}{\partial x\partial y} + D(1-2x)\frac{\partial^2}{\partial x\partial D} + D(1-2y)\frac{\partial^2}{\partial y\partial D} \\ &+ \frac{\theta}{4}(1-2x)\frac{\partial}{\partial x} + \frac{\theta}{4}(1-2y)\frac{\partial}{\partial y} - D(1 + \frac{\rho}{2} + \theta)\frac{\partial}{\partial D} \end{aligned}$$

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# Simplex

- Manipulate  $r^2$  instead of being clever with  $\mathcal{D}iff$
- Apply Linear Simplex Expansion on  $r^2$ 
  - Piecewise linear interpolation in high dimension
  - In every small region  $i$  of the tessellation:

$$S_i = a_{0i} + a_{1i}p_1 + a_{2i}p_2 + \dots + a_{9i}p_9$$

Where  $a_{0i}, \dots, a_{9i}$  are found by solving a series of 10-by-10 linear system

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Figure: I don't have a “Recipe”

# Lego

- Simplex expansion works like Lego

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- Instead of making an actual car, I am making a Lego car

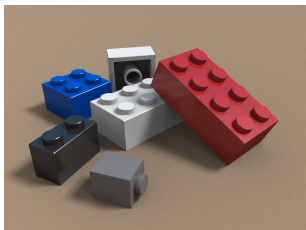


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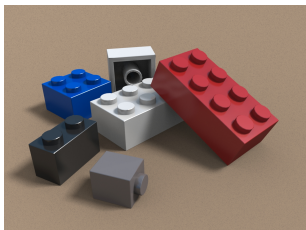


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Figure: Lego car

# Zoom



# Lego in Diff

- Lego bricks represents those **simple expectations**

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# Simplex and Lego

- Simplex expansion acts like a set of instruction, assembling and dismantling object, “Recipe” become irrelevant
- 

$$\begin{array}{rcl}
 \mathcal{D}iff[G_1] & \rightarrow & a_{0,1}^{g_1} \mathbb{E}[\mathbb{I}_1] + \cdots + a_{9,1}^{g_1} \mathbb{E}[\rho_9 \mathbb{I}_1] + \cdots + a_{0,k}^{g_1} \mathbb{E}[\mathbb{I}_k] + \cdots + a_{9,k}^{g_1} \mathbb{E}[\rho_9 \mathbb{I}_k] & = 0 \\
 \mathcal{D}iff[G_2] & \rightarrow & a_{0,1}^{g_2} \mathbb{E}[\mathbb{I}_1] + \cdots + a_{9,1}^{g_2} \mathbb{E}[\rho_9 \mathbb{I}_1] + \cdots + a_{0,k}^{g_2} \mathbb{E}[\mathbb{I}_k] + \cdots + a_{9,k}^{g_2} \mathbb{E}[\rho_9 \mathbb{I}_k] & = 0 \\
 & \vdots & \vdots & \\
 \mathcal{D}iff[G_{10k}] & \rightarrow & a_{0,1}^{g_{10k}} \mathbb{E}[\mathbb{I}_1] + \cdots + a_{9,1}^{g_{10k}} \mathbb{E}[\rho_9 \mathbb{I}_1] + \cdots + a_{0,k}^{g_{10k}} \mathbb{E}[\mathbb{I}_k] + \cdots + a_{9,k}^{g_{10k}} \mathbb{E}[\rho_9 \mathbb{I}_k] & = 0
 \end{array}$$

# Expression

- Final expression is in the form of a Padé approximation
  - Closely related to Taylor's expansion, but often superior when expanding a rational function
  - For  $k$  number of simplex division:

$$\mathbb{E}[r^2] = \frac{\alpha_0 + \alpha_1 N_e c + \alpha_2 (N_e c)^2 + \dots + \alpha_{10k} (N_e c)^{10K}}{\beta_0 + \beta_1 N_e c + \beta_2 (N_e c)^2 + \dots + \beta_{10k} (N_e c)^{10K}}$$

Where  $\alpha$  and  $\beta$  are coefficients and  $c$  is the recombination rate

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# Future work

- Numerically stable algorithm
- Completely Analytic derivation of the final expression by considering the limit as  $k \rightarrow \infty$

# Acknowledgment

- My Supervisor : Rachel Fewster
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You for listening!