

1.
 - I will talk about my thesis:
“Simulation of Simple Evolutionary System”
2.
 - I will give background, then address four questions, and make concluding remarks.
3.
 - Population is a collection of length l binary strings
 - A population can be represented by a vector; the j th component is the proportion of string j in the population.
 - \mathcal{R} is the set of length ℓ binary strings and operations under \mathcal{R} are bitwise modulo 2 operations.
4.
 - Crossover and mutation operators are defined using the bitwise operations in \mathcal{R} .
 - Crossover exchanges bits in parents u and v using crossover mask to produce children u' and v' using the rule $u' = um + v\bar{m}, v' = u\bar{m} + vm$
 - Mutation flips bits in x using mutation mask m according to rule x to $x + m$.
5.
 - This flowchart illustrates finite population genetic algorithm
 - Start from random initial population p .
 - Randomly select parents u and v
 - Crossover u and v to produce u' and v' according to randomly chosen crossover mask
 - Keep one of u', v' and mutate using randomly chosen mutation mask to produce gamete g
 - Repeat to form next generation $\tau(p)$
6.
 - In Random Heuristic Search,
 - Given population p , τ is transition rule that maps p to the next generation p' .

- The sequence shows finite population evolution which forms Markov chain
 - However, τ is stochastic function of crossover and mutation, and crossover masks and mutation masks are randomly chosen. So, $\tau(p)$ can not be predicted with certainty
 - χ_i is probability of choosing crossover mask i
 - μ_i is probability of choosing mutation mask i
- 7.
- In infinite population model, population is modeled as by a vector \mathbf{p}
 - \mathcal{G} is a function that maps \mathbf{p} to the next generation where $\mathcal{G}(p)_j$ is probability that string j occurs in the next generation.
 - The sequence shows evolution of \mathbf{p} under infinite population model.
 - The variance between next generation finite population and infinite population is given by expression where N is population size
- 8.
- Our first question concerns about convergence of finite population to infinite population
 - Chebyshev's inequality, Jensen's inequality and geometric point of view, all, suggests the distance might decrease as $1/\sqrt{r}$
 - But all of 3 implications are from inequalities. So the distance might decrease much smaller than $1/\sqrt{r}$.
 - We investigate if the distance infact decreases like $1/\sqrt{r}$ in practice. For this we set up simple diploid model.
- 9.
- It is infinite population model.
 - We consider diploid genome α with genome length l .
 - q is current generation population
 - q_α is prevalence of diploid α
 - $t_\alpha(g)$ is transmission function which is probability of gamete g being produced from parent α

- q' is next generation
- 10.
- Diploids can be determined in terms of haploid distributions.
 - Haploids can be determined in terms of diploid distributions.
 - Evolution equation can be expressed in terms of haploid distributions as

$$\mathbf{p}'_{\gamma_0} = \sum_{\alpha_0, \alpha_1} \mathbf{p}_{\alpha_0} \mathbf{p}_{\alpha_1} t_{\langle \alpha_0, \alpha_1 \rangle}(\gamma_0)$$

- In the matrix form, evolution equations can be expressed as
 - Where $M(g)$ is mixing matrix
- 11.
- This slide shows computation of crossover and mutation distributions, and transmission function
 - Transmission function is computed with this expression
 - There are 3 sum terms in it, which is expensive
- 12.
- W is walsh matrix
 - $\hat{w}(A)$ is walsh transform of matrix A , and \hat{w} is walsh transform of column vector w .
 - Mixing matrix is given by the expression

$$\widehat{M}_{u,v} = 2^{\ell-1} [uv = \mathbf{0}] \hat{\mu}_u \hat{\mu}_v \sum_{k \in \overline{u+v} \mathcal{R}} \chi_{k+u} + \chi_{k+v}$$

- And evolution eqn in Walsh basis is

$$\hat{\mathbf{p}}'_g = 2^{\ell/2} \sum_{i \in g \mathcal{R}} \hat{\mathbf{p}}_i \hat{\mathbf{p}}_{i+g} \widehat{M}_{i,i+g}$$

- Now if we compare evolution eqn in Walsh basis to what we had before, we got rid of matrix multiplication to compute next generation.
- We don't need to refer to 2^ℓ mixing matrices either, only one mixing matrix is required in walsh basis computation.
- Calculating each mixing matrix in previous eqn required 3 sum terms, which is reduced to 1 sum term in walsh basis.

13.
 - Reduction to haploid model and computing in Walsh basis simplified computations and made it efficient
 - 3 sum terms in transmission function is reduced to only one sum terms for computing mixing matrix in walsh basis
 - We only need one mixing matrix as opposed to 2^l
 - Consider $l = 14$, 2^{14} mixing matrices would require 32 TB of memory, while one mixing matrix would require only 2 GB of memory
 - We think that is great optimization in resource
14.
 - With naive computation as shown in equation (1), there would be $2^l \cdot 2^l$ multiplication terms to be computed
 - where f is finite population and q is infinite population
 - We simplify it as shown in eqn (2) where it needs only $2^l + \text{cardinal value of } S_f$ terms
 - S_f is set of diploids that are in finite populations
15.
 - We ran simulations with $\chi = 0.1$ and $\mu = 0.001$ for different values of l
 - These graphs show the results in logarithmic scale.
 - (point to axis) d is the distance, (point to axis) N is population size, (point to axis) n is number of generations
 - Data shows as population size increases, distance decreases and converge to infinite population
 - Graphs show flat surface so we model the data by linear equation
16.
 - Regression shows slope m is nearly equal to -0.5
 - This proves distance does decrease as $1/\sqrt{N}$
17.
 - $1/\sqrt{N}$ was theoretical upper bound but our simulaion shows it is really very good bound and the distance decreases as $1/\sqrt{N}$.

18. • Our 2nd question concerns about oscillation in finite populations.
19. • First I want to introduce term limit; we also call it fixed point.
 - The infinite population sequence $\mathbf{p}, \mathcal{G}(\mathbf{p}), \mathcal{G}^2(\mathbf{p}), \dots$ may converge to a point, and we call that point a fixed point.
 - But under certain conditions, the sequence converges to a periodic orbit between two fixed points, say \mathbf{p}^* and \mathbf{q}^*
20. • And the conditions are: For some g

$$\begin{aligned} -1 &= \sum_j (-1)^{g^T j} \mu_j \\ 1 &= \sum_{k \in \bar{g}\mathcal{R}} \chi_{k+g} + \chi_k \end{aligned}$$

- So our 2nd question is do finite populations also exhibit oscillations from random initial populations when infinite population oscillate?
21. • Akin, Hasting, Wright, Bidwell and Agapie have studied oscillations in the past.
 22. • Akin considered - continuous time model, we consider - discrete time model
 - Hastings' study - limited to two bits length, includes only crossover, not mutation
 - Wright and Bidwell compute a specific fitness function and a specific initial population for randomly generated mutation and crossover distributions in an attempt to find cyclic behavior
 - Wright and Agapie used dynamic mutation that depends upon where population is in the population space while we use static mutation
 - We study oscillation for
 - fixed fitness function and random: initial population, mutation and crossover distribution

- higher bit length (up to 14), and actually oscillation conditions for infinite populations are independent of string length
 - both haploid and diploid populations, and for both finite and infinite populations
 - We also visualize oscillation
- 23.
- Simulations were run for both haploid and diploid populations
 - for different string lengths l
 - and for population size 4096, 40960, 81920
 - To visualize oscillations, distances between fixed points (p^* , q^*) and population are plotted
- 24.
- Graphs show the results for haploid and diploid population of length 8
 - top haploid, bottom - diploids
 - Population size 4096, 40960, 81920 is in increasing order in columns
 - As population size increases, oscillation approaches the behavior exhibited by infinite population
 - We observe more randomness in diploids than in haploids for same string length and population size.
- 25.
- Graphs show the results for haploid and diploid population of length 12
 - We observe that increase in l degrades oscillation (particularly in diploids)
- 26.
- Our conclusions from this simulation are
 - Finite populations exhibit approximate oscillation when infinite populations oscillate
 - As population size increases, randomness decreases

27.
 - Question 3 concerns the robustness of finite population oscillation under violation in mutation
28.
 - We introduce violation ϵ as following
 - The modification in μ makes the Markov chain regular
 - This means no periodic orbits for finite population and no periodic orbits for infinite population
 - We study if finite population can exhibit approximate oscillations in that case
29.
 - Simulations were run for different values of ϵ , ℓ and N
 - Distances of population to limits p and q without violation are plotted
30.
 - Graphs show results for haploid population behavior of string length 8
 - Graphs are arranged in rows for increasing ϵ and columns for increasing population size
 - Graphs show convergence of finite population behavior to infinite population behavior as population increases
 - Results show oscillating behavior of population for smaller values of epsilon that diminishes with time.
 - Oscillation dies out for infinite population.
 - Even though finite population also appears to be dying out, since Markov chain is regular, finite population will oscillate infinitely often.
 - As value of epsilon grows, oscillation diminishes
31.
 - Graphs show results for diploid population behavior of string length 12
 - We observed oscillation also degrades as string length increases. It is seen in both haploid and diploid cases, but particularly noticeable in diploids.

- 32.
 - Finite populations exhibit approximate oscillation even if Markov chain is regular when violation is small
 - If violation becomes larger, finite population oscillation decreases
 - As string length increases, oscillation degrades
- 33.
 - Our last question concerns with the robustness of finite population oscillation under mutation- violation, our next question also concerns with the robustness of finite population but under violation in crossover
- 34.
 - Violation in crossover means no periodic orbit exists for infinite population
 - But we don't know if Markov chain is regular in this case
 - We investigate if finite population can exhibit approximate oscillations in this case
- 35.
 - We introduce violation ϵ as following
- 36.
 - Simulations were run for different values of ϵ, ℓ and N
 - Distances of population to limits p and q without violation are plotted
- 37.
 - Graphs show results for haploid population behavior of string length 8
 - Results show similar behavior to that of violation in mutation case
- 38.
 - Graphs show results for diploid population behavior of string length 12
 - However, we noticed that rate of damping of amplitudes of oscillation is slower than in mutation violation
 - And, we see more randomness in population behavior than in mutation violation, especially for diploid case
- 39.
 - Conclusions from experiment on violation in crossover are

- Finite populations exhibit approximate oscillation if violation is small
 - If violation becomes larger, finite population oscillation decreases
- 40.
- Overall conclusions from this research are
 - Vose's haploid model makes computation efficient in diploid case by reducing to haploid case
 - Distance between finite population and infinite population can decrease like $1/\sqrt{N}$
 - When infinite populations oscillate, finite populations exhibit approximate oscillation
 - Finite population exhibits approximate oscillation for small mutation violation
 - Finite populations exhibit approximate oscillation for small crossover violation