Markov Chain Modelling of Recombination Fraction of Chromosomes

Yuki Sugaya Ritei Shibata Keio University, Japan

> Shigeo Kamitsuji StaGen Co., Ltd.

Objective

- Markov Chain modelling of recombination fraction
- A new algorithm for linkage analysis

- "Probability inheritance" algorithm

Chromosomes in nuclear



http://www.tokyo-med.ac.jp/genet/chp.htm



Crossovers and Recombination



Recombination fraction (θ)

Probability of recombination between two loci

•Two loci are close



•Two loci are far



Map function $\theta(d)$

- *d* : genetic distance
 - the expected number of crossovers occurring on a single chromosome between two loci
- Haldane(1919),Kosambi(1944),Carter-Falconer(1951), Sturt(1976), Felsenstein(1979), Karlin(1984)
 - Example

Haldane(1919)
$$\theta(d) = \frac{1 - \exp^{-2d}}{d}$$

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Kosambi(1944)
$$heta(d) = rac{1}{2} anh 2d$$

Probabilistic model of recombination fraction –independent case-

Recombination fraction θ (n) between *n* loci

At most a crossover between the (*i-1*)-th and the *i*-th loci

$$X_{i} = \begin{cases} 1 & (\text{crossover between } i - 1 \text{ and } i) \\ 0 & (\text{no crossover between } i - 1 \text{ and } i) \end{cases} \qquad P(X_{i} = 1) = p_{0} \\ P(X_{i} = 0) = 1 - p_{0} \\ X_{i} \perp X_{j} \end{cases}$$

$$\theta(p_0, n) = P\left(\sum_{i=1}^n X_i = odd\right)$$
$$= \sum_{k:odd} \binom{n}{k} p_0^k (1-p_0)^{n-k}$$
$$= \frac{1}{2} \left\{ 1 - (1-2p_0)^n \right\}$$



Assumption

Markov chain modelling of recombination fraction -dependent case -

- Assumption
 - The direction of crossover :

the *0*-th locus \Rightarrow the *n*-th locus

Markov property of crossover

$$P(X_i = 1 | X_{i-1} = 0) = p_1 \qquad P(X_i = 1 | X_{i-1} = 1) = q_1$$
$$P(X_i = 0 | X_{i-1} = 0) = 1 - p_1 \qquad P(X_i = 0 | X_{i-1} = 1) = 1 - q_1$$

$$\theta = P\left(\sum_{i=1}^{n} X_i = odd\right)$$

complicated

Second order Markov modelling

$$Y_{i} = \begin{cases} 0 & \text{(paternal origin)} \\ 1 & \text{(maternal origin)} \end{cases} \begin{pmatrix} X_{i} = 0 \iff (Y_{i-1} = 0, Y_{i} = 0) \text{ or } (Y_{i-1} = 1, Y_{i} = 1) \\ X_{i} = 1 \iff (Y_{i-1} = 0, Y_{i} = 1) \text{ or } (Y_{i-1} = 1, Y_{i} = 0) \end{pmatrix}$$

$$p_{tu|st} = P(Y_i = t, Y_{i+1} = u | Y_{i-1} = s, Y_i = t)$$

No distinction between paternal and maternal origins

$$\begin{cases} p = p_{01|00} = p_{10|11} & 1 - p = p_{00|00} = p_{11|11} \\ q = p_{10|01} = p_{01|10} & 1 - q = p_{11|01} = p_{00|10} \end{cases} \begin{cases} P(Y_0 = 0) = P(Y_0 = 1) = \frac{1}{2} \\ P(Y_0 = 0, Y_1 = 0) = P(Y_0 = 1, Y_1 = 1) = \frac{1 - r}{2} \\ P(Y_0 = 0, Y_1 = 1) = P(Y_0 = 1, Y_1 = 0) = \frac{r}{2} \end{cases}$$

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Transition matrix

$$T = \begin{pmatrix} (00) & (10) & (01) & (11) \\ (00) & \left(\begin{array}{cccc} 1 - p & 0 & p & 0 \\ 1 - q & 0 & q & 0 \\ 0 & q & 0 & 1 - q \\ 0 & p & 0 & 1 - p \end{pmatrix}$$

Recombination fraction derived from Markov model

$$\theta(p,q,r,n) = P(Y_n = 1 | Y_0 = 0) \quad (= P(Y_n = 0 | Y_0 = 1))$$

$$= \sum_{k,l} \frac{P(Y_n = 1, Y_{n-1} = k, Y_1 = l, Y_0 = 0)}{P(Y_0 = 0)}$$

$$= \sum_{k,l} P(Y_n = 1, Y_{n-1} = k | Y_1 = l, Y_0 = 0) \frac{P(Y_1 = l, Y_0 = 0)}{P(Y_0 = 0)}$$

$$= \frac{1}{2} + \frac{1}{2(\lambda_1 - \lambda_2)} \left\{ (2r - 1)(\lambda_1^n - \lambda_2^n) + \lambda_1 \lambda_2 (\lambda_1^{n-1} - \lambda_2^{n-1}) \right\} \quad (n \ge 1)$$

$$\left(\lambda_1 = \frac{1 - p - q + \sqrt{(1 - p - q)^2 - 4(p - q)}}{2}, \quad \lambda_2 = \frac{1 - p - q - \sqrt{(1 - p - q)^2 - 4(p - q)}}{2}\right)$$

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$$(p = q = r)$$
 $\theta(p, n) = \frac{1}{2} \left\{ 1 - (1 - 2p)^n \right\}$
• $\theta(p, q, r, r) \to \frac{1}{2} (r, r) = 0$

•
$$\theta(p,q,r,n) \to \frac{-}{2} \quad (n \to \infty)$$

Pedigree data



disease locus (only phenotype is known)

Evaluation of the likelihood

Existing Algorithm

- Elston-Stewart algorithm(1971)

- backward algorithm
- Good for large pedigree and small number of markers
- Implemented in the LINKAGE program
- Lander-Green algorithm(1987)
 - Inheritance vector
 - Hidden Markov Model
 - Good for small pedigree and large number of markers
 - Implemented in the GENHUNTER program
- Our algorithm
 - "Probability inheritance" algorithm (Sugaya and Shibata,2005)
 - Good for large pedigree and large number of markers

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Phenotype to haplotype



- $A_{k,l} : a \text{ (affected) or } n \text{ (not affected)}$ $(H_{k,l}, H_{k,l+1}) \in \mathscr{H}^2 = \mathscr{H}^2_a \cup \mathscr{H}^2_n, \ \mathscr{H}^2_a \cap \mathscr{H}^2_n = \emptyset$ $A_{k,l} = a \Leftrightarrow (H_{k,l}, H_{k,l+1}) \in \mathscr{H}^2_a$ $A_{k,l} = n \Leftrightarrow (H_{k,l}, H_{k,l+1}) \in \mathscr{H}^2_n$ • $H_{k,l} : \text{Haplotype} \quad (H_{k,l} \in \mathscr{H})$
 - $\mathscr{A}_{k,l}$: A set of random variables that expresses affection status of an individual (k, l) and his ancestors

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Probability of phenotypes in a pedigree

$$P(\mathscr{A}_{1,1}) = \sum_{\mathscr{H}^2_{A_{1,1}}} P((H_{1,1}, H_{1,2}) = (h_{1,1}, h_{1,2}), \mathscr{A}_{2,1}, \mathscr{A}_{2,3})$$
$$= \sum_{\mathscr{H}^2_{A_{1,1}}} \left\{ P(H_{1,1} = h_{1,1}, \mathscr{A}_{2,1}) P(H_{1,2} = h_{1,2}, \mathscr{A}_{2,3}) \right\}$$



Probability inheritance



$$P(H_{k,l} = h_{k,l}, \mathscr{A}_{k+1,2l-1})$$

$$= \sum_{\mathscr{H}^{2}_{A_{k+1,2l-1}}} P(H_{k,l} = h_{k,l} \mid H_{k+1,2l-1} = h_{k+1,2l-1}, H_{k+1,2l} = h_{k+1,2l})$$

$$\cdot P(H_{k+1,2l-1} = h_{k+1,2l-1}, \mathscr{A}_{k+2,2(2l-1)-1}))$$

$$\cdot P(H_{k+1,2l} = h_{k+1,2l}, \mathscr{A}_{k+2,2(2l)-1})$$

Probability inheritance in three generations

$$P(H_{1,1} = h_1, \mathscr{A}_{2,1}) = \sum_{\mathscr{H}^2_{A_{2,1}}} P(H_{1,1} = h_1 \mid H_{2,1} = h_{2,1}, H_{2,2} = h_{2,2}) \times \left\{ \left\{ \sum_{\mathscr{H}^2_{A_{3,1}}} P(H_{2,1} = h_{2,1} \mid H_{3,1} = h_{3,1}, H_{3,2} = h_{3,2}) \times P(H_{3,1} = h_{3,1}) P(H_{3,2} = h_{3,2}) \right\} \times \left\{ \sum_{\mathscr{H}^2_{A_{3,3}}} P(H_{2,2} = h_{2,2} \mid H_{3,3} = h_{3,3}, H_{3,4} = h_{3,4}) \times P(H_{3,3} = h_{3,3}) P(H_{3,4} = h_{3,4}) \right\} \right\}$$



Example of "Probability inheritance"

maker allele : 1,2 disease allele : A,N

$$A_{k,l} = a \Leftrightarrow ((*, A), (*, A)) or ((*, A), (*, N))$$
$$A_{k,l} = n \Leftrightarrow ((*, N), (*, N))$$



Probability inheritance + Markov modelling of recombination fraction

$$H_{1,1} = (1, A), \ A_{1,1} = a$$

$$\mathbf{A} \stackrel{+}{+} \mathbf{A} \qquad \mathbf{A} \stackrel{+}{+} \stackrel{+}{+} \mathbf{N} \qquad \mathbf{N} \stackrel{+}{+} \stackrel{+}{+} \mathbf{A}$$

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$$P(H_{k,l} = (1, A)) = P_{1A}(k, l)$$
$$P(H_{k,l} = (*, A)) = P_{\bullet A}(k, l)$$
$$P(H_{k,l} = (1, *)) = P_{1\bullet}(k, l)$$

$$\theta(p,q,r,n) = \frac{1}{2} + \frac{1}{2(\lambda_1 - \lambda_2)} \left\{ (2r - 1)(\lambda_1^n - \lambda_2^n) + \lambda_1 \lambda_2 (\lambda_1^{n-1} - \lambda_2^{n-1}) \right\}$$

$$P(H_{1,1} = (1, A), A_{2,1} = a)$$

= $\frac{1}{2} (1 - \theta) \{ P_{1A}(2, 1) + P_{1A}(2, 2) \}$
+ $\frac{1}{2} \theta \{ P_{A}(2, 1)P_{1}(2, 2) + P_{1}(2, 1)P_{A}(2, 2) \}$

Summary

Markov Modelling of recombination fraction

$$\theta(p,q,r,n) = \frac{1}{2} + \frac{1}{2(\lambda_1 - \lambda_2)} \left\{ (2r - 1)(\lambda_1^n - \lambda_2^n) + \lambda_1 \lambda_2 (\lambda_1^{n-1} - \lambda_2^{n-1}) \right\}$$

- "Probability inheritance" algorithm
 - Likelihood can be evaluated by a simple recursive formula
 - Good for large pedigree and large number of markers

Markov modelling of recombination fraction + Probability inheritance

References

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Thank you for your listening.