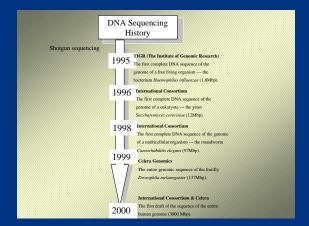
Dynamic Programming Algorithms for Haplotype Block Partitioning

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Sequencing History



Human Variations

Different kinds of human genome variation:

- Substitutions
- Insertions/Deletions
- Duplications
- Rearrangements

Substitutions of a single nucleotide represent 10% to 50% of human genome variation.

SNP = single nucleotide polymorphism

SNPs in the human genome about 1 in 600bps.





There are four haplotypes possible:



In general for n SNPs, there are 2^n possible haplotypes.

Genotype

Since we are diploid (have two chromosomes), things are more complex. Genotype of an individual:

It tells us that the individual has different base pairs on the maternal and paternal chromosome at site 1 and 2, the same base pairs at site 3.

Haplotypes of an individual may be:



Our Data Are from

Block of Limited Haplotype Diversity Revealed by High-resolution Scanning of Human Chromosome 21.

Patil N., Berno A.J., Hinds D.A., et al.

Science 294: 1719-1722 (23 Nov 2001).

Summary of the Data

- Sampled 23 ethnically diverse individuals
- Separated the 2 copies of chromosome 21 using rodent-human somatic cell hybrid technology
- 20 independent copies of chromosome 21 analyzed
- + 32.4×10^6 bases with 21.7×10^6 bases of unique sequence
- Essentially resequence using 3.4×10^9 oligonucleotide on $160~{\rm wafers}$
- Identity 35,989 SNPs
- Data at NCBI, dbSNP databases

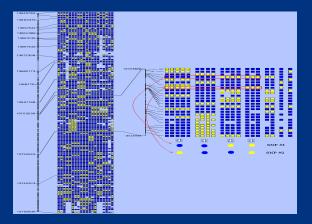
Haplotype Block Partitioning

• Objective:

To partition the haplotypes into blocks with minimum total number of SNPs required to account for most of the haplotype information in each block.

- Consecutive SNPs of size one or larger is a block only if the haplotypes represented more than once (common haplotypes) are more than α percent (e.g., α = 80%).
- Representative SNPs in a block are SNPs that can distinguish at least α percent (e.g., $\alpha = 80\%$) of haplotypes.

An Example of Haplotype Blocks



The Greedy Algorithm

- Finding all blocks with corresponding number of representative SNPs with a given percentage.
- Calculating the ratio with the total number of SNPs in the block to the number of representative SNPs.
- Selecting block with maximum ratio and discarding all other blocks with overlap with this block.
- Previous process is repeated in the remaining blocks until the blocks with no gaps and with every SNPs assigned to a block.
- The algorithm can not guarantee to minimize the number of representative SNPs.

The Dynamic Programming Algorithm

Develop a dynamic programming algorithm to partition the haplotypes into blocks to minimize the number of SNPs required to account for most of the haplotype information in each block.

Features of the program:

- Any measure of haplotype information can be used
- Guaranteed minimum number of representative SNPs
- Relatively fast

Mathematical Formulation

- The haplotypes are divided into blocks B_1, B_2, \ldots, B_I .
- Let $F(B_i)$ be the number of representative SNPs for block B_i .
- **Objective:** Minimize the total number of representative SNPs,

$$\sum_{i=1}^{I} F(B_i)$$

The Dynamic Programming Algorithm

- Let $block(r_i, r_{i+1}, ..., r_j) = 1$ if the SNPs $r_i, r_{i+1}, ..., r_j$ form a block, and 0 otherwise.
- Let S_j be the minimum number of representative SNPs for the optimal block partition of the first j SNPs, then

$$S_0 = 0$$

$$S_j = \min\{S_{i-1} + F(r_i, r_{i+1}, \dots, r_j),$$

if $block(r_i, r_{i+1}, \dots, r_j) = 1\}$

The Dynamic Programming Algorithm (cont.)

- First use the above dynamic programming algorithm to find S_n .
- Trace back to find the optimal block partition and the representative SNPs for the haplotypes.
- Finding the number of representative SNPs in a block is NP-hard. We use the enumeration method in this application.

Minimize the Number of Blocks

- The block partition with the minimum number of SNPs is not unique.
- Choose that partition with the minimum number of blocks. Let C_j be the minimum number of blocks requiring S_j in the first j SNPs,

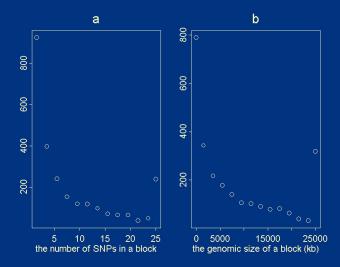
 $C_0 = 0$ $C_j = \min\{C_{i-1} + 1, \text{if } block(r_i, r_{i+1}, \dots, r_j) = 1, \\ \text{and } S_j = S_{i-1} + F(r_i, r_{i+1}, \dots, r_j)\}$

Results

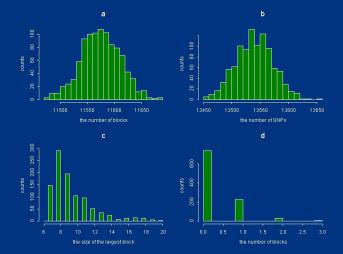
Block partition using dynamic programming vs greedy algorithm ($\alpha = 80\%$):

			size of	% blocks
	number	number	largest	with > 10
	of SNPs	of blocks	block	SNPs
dynamic	2,575	3,582	128	28.8
greedy	4,536	4,135	114	14.2
difference	37.7%	21.5%	12.3%	102.8%

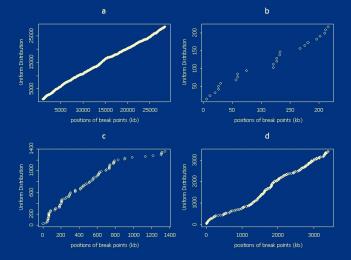
The number of blocks as a function of the size of the blocks



Testing the significance of the results using permutation tests



Homogeneity of the block break points



Limited Budget Haplotype Block Partitioning

• Objective:

To partition the haplotypes into blocks with a certain number of representative SNPs that can cover maximum length of genome.

- Consecutive SNPs of size one or larger is a block only if the haplotypes represented more than once (common haplotypes) are more than α percent (e.g., α = 80%).
- Representative SNPs in a block are SNPs that can distinguish at least α percent (e.g., $\alpha = 80\%$) of haplotypes.

Mathematical Formulation

- Let N be the number of representative SNPs required.
- The haplotypes are divided into blocks B_1, \ldots, B_I and deleted intervals D_1, \ldots, D_J .
- Let $F(B_i)$ be the number of representative SNPs for block B_i .
- Let $L(r_i, \ldots, r_j)$ be the length from SNPs r_i to SNPs r_j .

 $L(\cdot)$ can be the number of SNPs or actual genomic length.

Mathematical Objective

• Objective: Maximize the covered length of genome:



under the constraint:

$$\sum_{i=1}^{I} F(B_i) \le N.$$

• If N is equal or larger than the minimum number of total representative SNPs, the block partition has no deleted intervals.

Two Dimensional Dynamic Programming Algorithm

- Let $block(r_i, r_{i+1}, ..., r_j) = 1$ if the SNPs $r_i, r_{i+1}, ..., r_j$ form a block, and 0 otherwise.
- Let $S_{j,k}$ be the maximum length of genome that is covered by k representative SNPs for the optimal block partition of the first j SNPs. We set $S_{0,k} = 0$ for any $k \ge 0$ and $S_{0,k} = -\infty$ for any k < 0, then

$$S_{j,k} = \max \left\{ \begin{array}{l} \max S_{i-1,k-F(r_i,...,r_j)} + L(r_i,...,r_j) \\ \text{if } block(r_i,r_{i+1},...,r_j) = 1, \\ S_{j-1,k} \end{array} \right\}$$

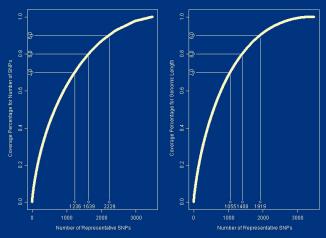
The Two Dimensional Dynamic Programming Algorithm (cont.)

- First use the above dynamic programming algorithm to find $S_{n,N}$.
- Trace back to find the optimal block partition.
- If there is $1 \le i^* < j$, s.t. $block(r_{i^*}, \dots, r_j) = 1$ and $S_{j,k} = S_{i^*-1,k-F(r_{i^*},\dots,r_j)} + L(r_{i^*},\dots,r_j)$

then this block $(r_{i^*}, r_{i^*+1}, \ldots, r_j)$ is retained in the partition. Otherwise, $S_{j,k} = S_{j-1,k}$ and the SNPs j is deleted.

Results by the two dimensional dynamic programming algorithm

- Set α = 0.80 to define the block and find the number of representative SNPs.
- The number of representative SNPs, $F(B_i)$, is set to 1 if $F(B_i) = 0$.
- The scatter plot for the number of representative SNPs and the coverage percentage: (a)L(·) is the number of SNPs; (b) L(·) is the actual genomic length.



а

b

The Parametric Dynamic Programming Algorithm

- Let $block(r_i, r_{i+1}, ..., r_j) = 1$ if the SNPs $r_i, r_{i+1}, ..., r_j$ form a block, and 0 otherwise.
- Given a deletion parameter β, let S_j(β) be the minimum score for the optimal block partition of the first j SNPs. We set S₀(β) = 0, then:

$$S_j(\beta) = \min \left\{ \begin{array}{l} \min S_{i-1}(\beta) + F(r_i, \dots, r_j) \\ \text{if } block(r_i, r_{i+1}, \dots, r_j) = 1, \\ \min_{1 \le i \le j} S_{i-1}(\beta) + \beta \cdot L(r_i, \dots, r_j) \end{array} \right\}.$$

The Parametric Dynamic Programming Algorithm (cont.)

- First use the above dynamic programming algorithm to find S_n(β) for a given β.
- Trace back to find the optimal block partition.
- If there is $1 \le i^* < j$, s.t. $block(r_{i^*}, \ldots, r_j) = 1$ and $S_j(\beta) = S_{i^*-1}(\beta) + A(r_{i^*}, \ldots, r_j)$

then this block $(r_{i^*}, r_{i^*+1}, \ldots, r_j)$ is retained in the partition. Otherwise, there is $1 \leq i^* < j$ s.t. $S_j(\beta) = S_{i^*-1}(\beta) + \beta \cdot L(r_{i^*}, \ldots, r_j)$ and this interval $(r_{i^*}, r_{i^*+1}, \ldots, r_j)$ is deleted.

The Properties of $S_n(\beta)$

 Algorithm. The computation of S_j(β) can be reduced to:

$$S_j(\beta) = \min \left\{ \begin{array}{l} \min S_{i-1}(\beta) + F(r_i, \dots, r_j) \\ \text{if } block(r_i, r_{i+1}, \dots, r_j) = 1, \\ S_{i-1}(\beta) + \beta \cdot L(r_j, \dots, r_j) \end{array} \right\}.$$

• **Properties.** $S_n(\beta)$ is increasing, piecewise linear and convex. The right-most linear segment of $S_n(\beta)$ is constant. The intercept and slope for $S_n(\beta)$ are the total number of representative SNPs and the total length of deleted intervals, respectively.

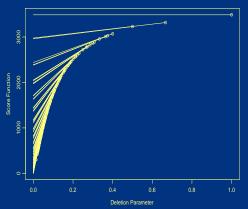
The Algorithm for Finding $S_n(\beta)$

- 1. Set i=1 and let L_i and L_{∞} be lines defined by $S_n(0)$ and $S_n(\infty)$ respect to β , respectively.
- 2. Find the intersection point (x_i, y_i) for L_i and L_{∞} and calculate $S_n(x_i)$. If $S_n(x_i) = y_i$, we know the entire function of $S_n(\beta)$. Otherwise, $S_n(x_i) > y_i$.
- 3. Find the line that through $(x_i, S_n(x_i))$ and continue intersections for this line with the line L_i until we find the intersection point (x_i, y_i) satisfies $S_n(x_i) = y_i$.
- 4. Take the line segment of $S_n(\beta)$ found just before L_i as L_{i+1} .
- 5. Set i = i + 1 and repeat 2-4.

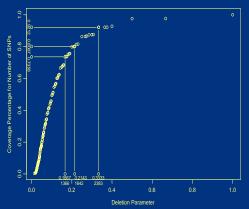
Results by the parametric dynamic programming algorithm

- $L(\cdot)$ is the number of SNPs.
- Set α = 0.80 to define the block and find the number of representative SNPs.
- The number of representative SNPs, $F(B_i)$, is set to 1 if $F(B_i) = 0$.
- The plot for the function $S_n(\beta)$.
- The scatter plot for (a) the deletion parameter and the the coverage percentage; (b) the number of representative SNPs and the coverage percentage.

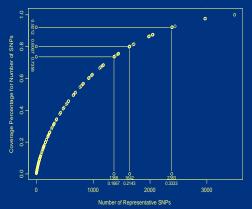
The slope of parametric dynamic programming



The slope of parametric dynamic programming



The slope of parametric dynamic programming



Statistics for different deletion parameters

Statistics for some deletion parameters with $\alpha = 0.80$. $F(B_i)$ is at least 1 and $L(\cdot)$ is the number of SNPs.

			Number	Number	Number
	Deletion	Coverage	of	of Deleted	of Rep.
	Parameters	Percentage	Blocks	Intervals	SNPs
1	0.16667	0.735806	748	626	1366
2	0.18182	0.752335	770	641	1431
3	0.21429	0.800733	896	706	1642
4	0.25000	0.863141	1071	748	1974
5	1.00000	1.00000	1696	0	3488

Comparison of block partitions for different deletion parameters

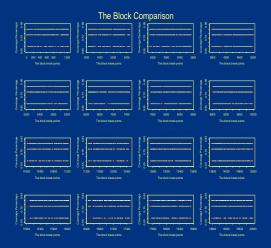
The number of identical blocks for two deletion parameters with $\alpha = 0.80$. $F(B_i)$ is at least 1 and $L(\cdot)$ is the number of SNPs.

Deletion					
Parameters	0.16667	0.18182	0.21429	0.25000	1.0000
0.16667	-	736	682	588	224
0.18182	736		715	617	234
0.21429	682	715		770	302
0.25000	588	617	770		412
1.00000	224	234	302	412	

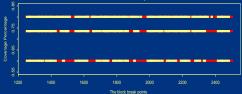
Comparison of deleted intervals for different deletion parameters

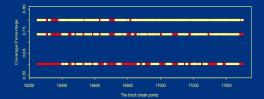
The number of identical deleted intervals for two deletion parameters with $\alpha = 0.80$. $F(B_i)$ is at least 1 and $L(\cdot)$ is the number of SNPs.

Deletion					
Parameters	0.16667	0.18182	0.21429	0.25000	1.0000
0.16667	-	592	439	285	0
0.18182	592		479	313	0
0.21429	439	479		452	0
0.25000	285	313	452		0
1.00000	0	0	0	0	0

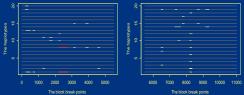


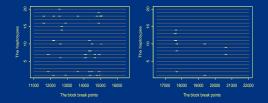
The Block Comparison





The Local Block Structure





Summary

- Develop a dynamic programming algorithm for haplotype block partition to minimize the number of representative SNPs.
- Develop a two dimensional dynamic programming and a parametric dynamic programming algorithm to maximize the covered length of genome with a certain number of representative SNPs.
- Apply the algorithm to the Chromosome 21 data. Test the statistical significance of the block partition results.
- Regions of biological interest can be identified based on the block break points.

Perspectives

- Faster algorithms to find the number of representative SNPs in a block may be needed for large scale problems
- How many chromosomes are needed to capture the common haplotype structures of the general populations?
- What are the biological reasons for the observed haplotype structure?
- For association studies, how much information can be lost (or gained) by using only the representative SNPs instead of all the SNPs?

Acknowledgements

- Thank Perlegen for making the data available and their excellent ideas on formulating the problem.
- USC computational biology group for many interesting ideas and discussions.
- Supported by NIH and NSF.