

## Appendix

### 1 Proof of Lemma 2

**Proof 1** Let  $\{(f_1, f_2), (f_2, f_3), \dots, (f_{k-1}, f_k), (f_k, f_1)\}$  be the edges of a simple cycle  $c_s$  in  $G_F$  of length  $k$  fragments (vertices). We can partition the fragments into two sets such that each set corresponds to the haplotypes of the individual. If  $k$  is even, then we can partition the even fragments  $(f_2, \dots, f_k)$  and odd fragments  $(f_1, \dots, f_{k-1})$  into two sets such that each set does not contain internal fragment conflicts. Likewise, if  $k$  is odd, then no such partition exists because  $f_k$  conflicts with  $f_1$  and  $f_{k-1}$ . The function that takes a cycle in  $G_C$  and computes the number of  $\frac{01}{10}$  (negative) edges is denoted  $\text{neg}()$ . We claim that for  $k$  even,  $\text{neg}(c_s)$  is even and for  $k$  odd  $\text{neg}(c_s)$  is odd. For this proof we consider any length  $k - 1$  subset of vertices in  $c$  and without loss of generality we assume this subset is  $v_1, \dots, v_{k-1}$ . Consider any two adjacent fragments in this cycle  $f_i$  and  $f_j$  such that  $i < j$  and they share the  $k^{\text{th}}$  SNP. As we iterate through fragments of the cycle, we call the allele that will be paired with the next fragment the active allele. If  $(s_k, s_{k+1}) < 0$  then  $f_{j,k+1} = f_{i,k}$ , that is, the active allele that will pair with  $f_{j+1}$  is the same allele as  $f_{i,k}$ . However, if  $(s_k, s_{k+1}) > 0$  then  $f_{j,k+1} \neq f_{i,k}$ , and the active allele that will pair with  $f_{j+1}$  will be the opposite allele as  $f_{i,k}$ . Thus negative edges in  $G_C$  do not change the active allele while positive edges in  $G_C$  flip the active allele from 0 to 1 (or vice-versa).

Case (1):  $k$  even. The  $v_1, \dots, v_{k-1}$  subset either has an even or odd number of negative pairwise phase relationships. Case 1.a: Even number of negative pairwise phase relationships; odd number of positive pairwise phase relationships. The active allele of  $v_{k-1}$  is the same as the active allele of  $v_1$  therefore  $v_k$  must be induce a positive pairwise phase relationship. Case 1.b: Odd number of negative pairwise phase relationships; even number of positive pairwise phase relationships. The active allele of  $v_{k-1}$  is different from the active allele of  $v_1$  therefore  $v_k$  must be induce a negative pairwise phase relationship. In both cases 1.a and 1.b the total number of negative edges is even.

Case(2):  $k$  is odd. Case 2.a: Even number of negative pairwise phase relationships; even number of positive pairwise phase relationships. The active allele of  $v_{k-1}$  is different from the active allele of  $v_1$  therefore  $v_k$  must be induce a negative pairwise phase relationship. Case 2.b: Odd number of negative pairwise phase relationships; odd number of positive pairwise phase relationships. The active allele of  $v_{k-1}$  is the same as the active allele of  $v_1$  therefore  $v_k$  must be induce a positive pairwise phase relationship. In both cases 2.a and 2.b the total number of negative edges is odd.

### 2 MWVR Proof

Theorem: MWVR is NP-hard.

**Proof 2** The reduction is from the problem of removing the minimum number of edges of a graph to make it bipartite. Let  $G$  be an arbitrary graph and  $M$  the SNP-fragment matrix as defined in Lemma 1 which encodes the fragment conflict graph  $G_F = G$ .  $G_F$  may contain a number of cycles of odd length which produce conflicting cycles in the compass graph  $G_C$  by Lemma 2. Each vertex in  $G_C$  corresponds to an edge in  $G_F$  by Lemma 1. The vertex set solution to the MVR optimization  $L$  yields the minimum number of vertices required to remove all of the conflicting cycles in  $G_C$ . Because a graph is bipartite if and only if it contains no odd length cycles and  $G_C$  is the line graph of  $G_F$ , the removal of these vertices corresponds to removal of edges; the minimum of which makes  $G_F$  bipartite.

### 3 Pacific Biosciences run times

	HapCompass MWER	HapCompass MEC	HapCUT	Levy
avg. time (s)	10	10.8	13.6	19.3
avg. memory (MB)	1251	1489	43.2	1049

Table 1: Average resource requirements for PacBio haplotype assembly runs. The HapCompass software is not optimized for minimal memory usage which is exemplified in the memory requirement results of the Levy *et al.* (2007) algorithm. This algorithm is implemented within the HapCompass software and should have a very small fingerprint but requires about a gigabyte of memory. Reducing the input fragment set into a secondary format prior to haplotype assembly (HapCUT does this) reduces our memory footprint by a factor of 10-100 times.

### 4 1000 Genomes Project Results

Chr	MWER			MEC			Levy			HapCUT		
	FMPR	BFM	MEC	FMPR	BFM	MEC	FMPR	BFM	MEC	FMPR	BFM	MEC
1	<b>3421</b>	<b>2348</b>	<b>2371</b>	3681	2519	2545	3619	2594	2632	3520	2423	2441
2	<b>4891</b>	<b>2930</b>	<b>2996</b>	5193	3081	3166	5154	3175	3273	5140	3022	3072
3	<b>3696</b>	<b>2394</b>	<b>2449</b>	4014	2585	2629	3823	2643	2703	3789	2476	2511
4	<b>4846</b>	<b>2710</b>	<b>2777</b>	5136	2906	2976	4891	2899	2974	4971	2805	2846
5	<b>3569</b>	<b>2245</b>	<b>2265</b>	3847	2428	2451	3851	2581	2606	3650	2290	2299
6	10425	<b>3603</b>	4032	10944	3846	4265	<b>9468</b>	3700	4075	10597	3630	<b>4030</b>
7	<b>3512</b>	<b>2138</b>	<b>2173</b>	3768	2288	2330	3677	2358	2407	3621	2214	2238
8	<b>2894</b>	<b>1864</b>	<b>1891</b>	3142	1999	2029	3048	2084	2118	2979	1947	1951
9	2844	<b>1551</b>	<b>1572</b>	3039	1667	1689	<b>2737</b>	1655	1687	2884	1580	1591
10	<b>2743</b>	<b>1857</b>	<b>1875</b>	2952	1981	2001	2838	2027	2048	2836	1932	1940
11	2662	<b>1634</b>	<b>1650</b>	2837	1727	1749	<b>2643</b>	1739	1778	2728	1694	1693
12	<b>2620</b>	<b>1627</b>	<b>1657</b>	2833	1784	1811	2786	1819	1856	2676	1678	1687
13	2503	<b>1461</b>	<b>1477</b>	2625	1554	1573	<b>2473</b>	1558	1576	2548	1490	1501
14	<b>1442</b>	<b>1020</b>	<b>1027</b>	1525	1070	1079	1512	1094	1102	1471	1045	1044
15	<b>1635</b>	<b>1085</b>	<b>1097</b>	1786	1168	1189	1757	1254	1272	1696	1133	1142
16	<b>2158</b>	<b>1308</b>	<b>1344</b>	2297	1410	1435	2198	1405	1458	2205	1333	1368
17	2797	1219	1320	3099	1354	1460	<b>2493</b>	1230	1305	2788	<b>1216</b>	<b>1299</b>
18	<b>1457</b>	<b>982</b>	<b>985</b>	1629	1088	1094	1563	1118	1130	1490	1013	1009
19	<b>1292</b>	<b>803</b>	<b>815</b>	1404	865	879	1369	901	918	1324	816	826
20	<b>1169</b>	<b>808</b>	<b>817</b>	1247	859	866	1279	924	939	1210	846	847
21	<b>871</b>	<b>545</b>	<b>558</b>	916	581	588	912	589	601	901	563	574
22	<b>681</b>	<b>446</b>	<b>449</b>	709	461	465	698	485	488	700	460	463

Table 2: Results of the NA12878 1000 Genomes Project 454 haplotype assemblies for chromosomes (chr) 1-22 and algorithms HapCompass MWER, HapCompass MEC, Levy *et al.* (2007), and HapCUT.

### References

Levy, S., Sutton, G., *et al.* (2007). The diploid genome sequence of an individual human. *PLoS biology*, **5**(10), e254.