

Gene Verification and Discovery by Walking Tree Method

Tai Hsu, Paul Cull

Computer Science Department, Oregon State University

Corvallis, OR 97331

(hsuta@cs.orst.edu, pc@cs.orst.edu)

The Walking Tree Method [3, 4, 5, 18] is an approximate string alignment method that can handle insertions, deletions, substitutions, translocations, and more than one level of inversions all together. Moreover, it tends to highlight gene locations, and helps discover unknown genes. Its recent improvements in runtime and space use extends its capability in exploring large strings. We will briefly describe the Walking Tree Method with its recent improvements [18], and demonstrate its speed and ability to align real complete genomes such as *Borrelia burgdorferi* (910724 base pairs of its single chromosome) and *Chlamydia trachomatis* (1042519 base pairs) in reasonable time, and to locate and verify genes.

1. Introduction

Most biological string matching methods are based on the edit-distance model [15]. These methods assume that changes between strings occur locally. But, evidence shows that large scale changes are possible [7]. For example, large pieces of DNA can be moved from one location to another (translocations), or replaced by their reversed complements (inversions). Schöniger and Waterman [14] extended the edit-distance model to handle inversions, but their method handled only one level of inversion. Hannenhalli's algorithm [10] for the "translocation" problem runs in polynomial time, but it requires gene locations to be known. Furthermore, it seems unlikely that any simple model will be able to capture the minimum biologically correct distance between two strings. In all likelihood finding the fewest operations that have to be applied to one string to obtain another string will probably require trying all possible sequences of operations. Trying all possible sequences is computationally intractable. This intractability has been confirmed by a recent proof by Caprara [2] that determining the minimum number of flips needed to sort a sequence is an NP-complete problem. Although signed flips can be sorted in polynomial time [11], apparently, we need a method that can handle insertions, deletions, substitutions, translocations, and inversions together. The Walking Tree heuristic handles translocations and multi-level inversions well, and also tends to highlight genes [3, 4, 5, 18].

2. Walking Tree Method

2.1 The Method

The problem is to find an approximate biologically reasonable alignment between two strings, one called pattern P, and the other called text T. Our metaphor is to consider the data structure as a walking tree with $|P|$ leaves, one for each character in the pattern. When the walking tree is considering position $l + 1$, the internal nodes remember some of the information for the best alignment within the first l characters of the text (Figure 1). On the basis of this remembered information and the comparisons of the leaves with the text characters under them, the leaves update their information and pass this information to their parents. The data will percolate up to the root where a new best score is calculated. The tree can then walk to the next position by moving each of its leaves one character to the right. The whole text has been processed when the leftmost leaf of the walking tree has processed the rightmost character of the text.

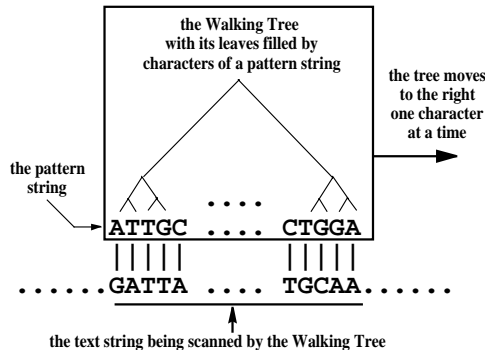


Figure 1: This picture shows the walking tree's structure, a binary tree. Leaves of the tree contain the characters of the pattern string P. After comparing each leaf with a corresponding character of the text string, the walking tree updates its nodes with new scores, then moves to the next position by moving each of its leaves one character to the right. Then it repeats the leaf comparison, and updates its node scores until it reaches the end of the text string.

To define a scoring system that captures some biological intuitions, we use a function that gives a positive contribution based on the similarity between aligned characters, and a negative contribution that is related to the number and length of gaps, translocations, and inversions. A gap in an alignment occurs when adjacent characters in the pattern are aligned with non-adjacent characters in the text. The length of the gap is the number of characters between the non-adjacent characters in the text. The detailed description of the resource usage of the method can be found in Cull, Holloway and Hsu's papers [3, 4, 5, 18]

2.2 Improvements in Speed and Space

The binary tree structure of the Walking Tree makes it extremely easy to implement a parallel version (Figure 2). Furthermore, inexpensive vector processors can be used because each node of the tree does the same operations at each scanning position. Each parent node of the walking tree simultaneously updates its score and position whenever it observes a better score.

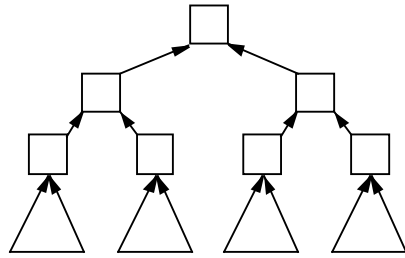


Figure 2: This picture shows the parallelization of the walking tree method. Given one processor per node of the tree, each child sends its current information to its parent; so a parent can update its best score and position by the information. Since the tree is $\log_2|P|$ high, $\Theta(\log_2|P|)$ startup time is needed for the root to receive its first information from leaves. After the startup time, all nodes work simultaneously; so, each text scan step takes $\Theta(1)$ time. The parallel runtime is $\Theta(\log_2|P| + |T|)$, i.e., $\Theta(|T|)$ because $|T| \geq |P|$.

We recognized that the alignment copying in the original design [3, 4, 5] was passively activated whenever a better score occurred. It's better to postpone the copying to allow faster scoring at the tree nodes. Based on this idea, we discovered improvements [18] for both the sequential and the parallel versions of the Walking Tree Method by using a state-caching technique similar to that used in recovering from program crashes (Figure 5.)

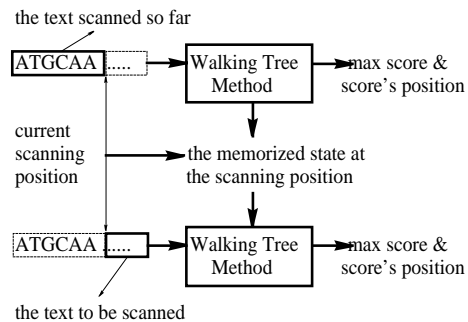


Figure 5: We use a technique similar to recovering a crashed program by saving its state before crashes. The memorized states record the states of the walking tree and the corresponding scanning positions of the text string. Once we have the recorded information, we can scan the text from the position we have memorized to avoid scanning from the first position of the text.

The improved sequential version [18] of the Walking Tree Method guarantees $\Theta(|P|*|T|^k)$ runtime using $\Theta(|P|*(\log_2|P|)^{1/k})$ space. With $\Theta(|P|)$ CPUs, the improved parallel version [18] guarantees $\Theta(|T|)$ runtime using $\Theta(|P|*\log_2|P|)$ space by reducing inter-processor communication to make CPUs spend more time on working rather than talking to each other. The improvements [18] also allows us to use a simpler implementation to overlap communication and computation in a shared memory model, e.g., a cluster of network computers. Exploring large strings becomes feasible. Fig. 14 shows the result of the new improvement versus the original method (the parallelization uses MPICH (version 1.1.0) [9, 12] and a cluster of Intel Pentium II 300 MHz machines (running Red Hat Linux 5.2) connected by a 100 Mbps switch). Our model doesn't consider the PRAM model [8] because the PRAM model [8] is considered unrealistic [1, 6], in that it assumes unlimited bandwidth and free interprocessor communication.

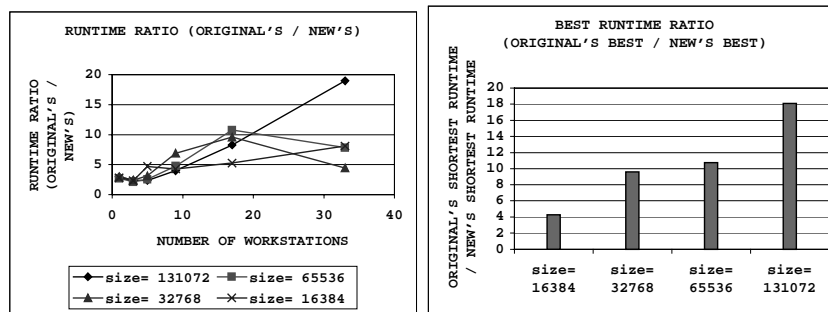


Figure 14: As the first picture shows, when only 1 workstation is used, the new method is about 2.7 times as fast as the original one; therefore, we should see the same speed ratio when using more workstations, if both methods are equally good in network parallelization. As the first picture shows, both are equally good when 5 workstations or less are used, but the new method prevails when 9 workstations or more are used. In addition, if both are equally good in network parallelization, the ratios of their best (i.e., the shortest) runtimes should be around 2.7 as well. That is, each method can use any number of the 33 workstations to get the best result for a particular input size. However, as the second picture shows, the new method prevails constantly with speed ratios better than 2.7, especially when the input size = 131072.

3. Previous Result

Our previous result showed that the Walking Tree can detect unknown genes, and align translocations and inversions. In Figure 15 and Figure 16, we show two alignments of two pairs of real DNA sequences. They are identical to the alignments found in Cull *et al*'s paper [5, 18].

4. New Result

With our recent improved Walking Tree Method, we are now able to align two real complete genomes (Fig 17, Fig 18, Fig 19), *Borrelia burgdorferi* [16] (910724 base pairs of its single chromosome and *Chlamydia trachomatis* [17] (1042519 base pairs) in 24 hours using 65 Pentium II 300MHz PC's. We separate the new result into 3 categories:

1. Matched regions that have annotations on both DNAs (TABLE A, TABLE B, Fig 17)
2. Matched regions that have annotations on only one DNA (Fig 18)
3. Matched regions that have no annotations on either DNA (Fig 19)

There are 103 matches in category 1, i.e., 40 translocations and 63 inversions. There are 148 matches in category 2, i.e., 86 translocations and 62 inversions. There are 1367 matches in category 3, i.e., 700 translocations and 667 inversions.

TABLE A: TRANSLOCATIONS									
Borrelia burgdorferi				Chlamydia trachomatis					
Aligned positions by Walking Tree Method		Gene annotations & locations from Genbank		Aligned positions by Walking Tree Method		Gene annotations & locations from Genbank			
88704	92799	89200	89814	BB0092	342373	346480	342872	343483	atpD
		89811	91115	BB0093			343468	344784	atpB
		91137	92792	BB0094			344787	346562	atpA
549888	551935	549642	551723	BB0540	505723	507776	505508	507592	fusA
454688	456702	454484	456403	BB0436	213601	215655	212937	215351	gyrB_1
							215354	215704	CT191
588032	589823	588066	589667	BB0575	204388	206112	204429	206048	pyrG
200960	202494	201052	202578	BB0201	299917	301427	300027	301478	murE
84672	86015	84041	85720	BB0088	75274	76600	74661	76469	lepA
456960	458239	456576	458036	BB0437	306732	307968	306433	307800	dnaA_2
114944	116223	114807	115508	BB0117	340937	342196	340917	342866	atpI
863488	864767	863636	865042	BB0817	897002	898255	897403	897822	CT763
686080	687103	685977	686507	BB0647	997281	998185	997122	997640	CT847
							997656	998162	CT848
326656	327678	326699	327757	BB0322	807860	808838	807691	808218	CT702
797696	798719	798057	799016	BB0755	82926	83736	82824	83780	ytgD
354816	355839	354648	355298	BB0346	716792	717677	717087	717770	cpxR
866304	867326	866494	866694	BB0820	898962	899997	898940	899272	rsbV_2
		866681	867601	BB0821			899276	900295	miaA
		8412	9197	BB0008	514258	515301	514382	514606	CT444.1
8704	9727	673342	674778	BB0636	207123	208166	206802	208121	zwf
673792	674815	735343	736686	BB0694	30326	31339	29938	31284	ffh
735744	736767	526325	527305	BB0515	115999	117014	115919	116974	trxB
526337	527359	769547	771145	BB0730	430604	431626	430608	432185	pgi
769537	770559	345063	346364	BB0337	661878	662880	661850	663124	eno
345088	346111	752134	753219	BB0715	818326	819402	818358	819458	mreB
752128	753151	235595	237142	BB0230	566490	567573	566631	568025	rho
235520	236543	817393	817956	BB0776	997263	998285	997122	997640	CT847
817153	818175						997656	998162	CT848
		317247	318026	BB0309	306526	307394	306433	307800	dnaA_2
317440	318335	318119	318256	BB0310					
		586212	587024	BB0573	827709	828882	828429	828794	CT716
586496	587390	759586	760215	BB0721	987691	988495	987715	988779	CT839
759296	760063	528104	529198	BB0517	389889	390701	389567	390745	dnaJ
528384	529151	294785	295228	BB0284	690685	691202	690426	691121	CT610
294912	295423	60036	60554	BB0065	303646	304156	303731	304018	CT271
60160	60670	512393	513148	BB0507	828414	828982	828429	828794	CT716
512513	513022	459525	459824	BB0439	995630	996100	995570	996061	yfhC
459264	459775	823167	823811	BB0786	995596	996039	995570	996061	yfhC
823297	823807	738851	738964	BB0700	995494	996068	995570	996061	yfhC
738560	739006	531851	531967	BB0520	828452	828707	828429	828794	CT716
531712	531967								

Table B: INVERSIONS							
Borrelia burgdorferi				Chlamydia trachomatis			
Aligned positions by Walking Tree Method		Gene annotations & locations from Genbank		Aligned positions by Walking Tree Method		Gene annotations & locations from Genbank	

Pacific Symposium on Biocomputing 6:287-298 (2001)

500355	497028	497245	497874	BB0479	592489	595884	592462	593136	rs3
		497880	498191	BB0480			593146	593481	r122
		498213	499046	BB0481			593500	593766	rs19
		499056	499334	BB0482			593772	594626	r12
		499341	499703	BB0483			594650	594985	r123
		499707	500588	BB0484			595001	595669	r14
438403	435204	435201	435312	5S_rrlA	877808	881136	878039	880902	23SrRNA_2
		435334	438267	23S_rrlA			881027	881143	5SrRNA_2
441731	438660	438590	441508	23S_rrlB	877746	880813	878039	880902	23SrRNA_2
884099	882052	881085	884213	BB0833	21536	23573	21432	24542	ileS
258434	256452	256463	258985	BB0251	236302	238242	235766	238225	leuS
531331	529540	529198	531105	BB0518	451380	453186	451614	453596	dnaK
639875	638085	637963	638556	BB0611	811081	812898	811130	812389	clpX
		638580	639872	BB0612			812399	813010	clpP_2
446083	444548	444581	446118	16S	876201	877753	876174	877723	16SrRNA_2
467330	465796	465518	467038	BB0446	986947	988305	986612	987712	CT838
502659	501124	501215	501469	BB0487	590356	591861	590272	590814	r15
		501491	501865	BB0488			590816	591151	r124
		501880	502185	BB0489			591164	591532	r114
		502191	502739	BB0490			591549	591800	rs17
690051	688516	688490	690127	BB0649	126399	127939	126336	127970	groEL_1
536963	535621	535704	537527	BB0526	340342	341750	340429	340875	atpK
496259	494980	495012	496217	BB0476	362055	363207	361980	363164	tufA
350850	349573	349600	351090	BB0342	2295	3565	2108	3583	gatA
180867	179588	179540	181423	BB0178	577486	578744	576941	578723	gidA
370051	369028	368885	370027	BB0361	828101	828995	828429	828794	CT716
803715	802692	802838	803212	BB0760	325245	326318	325478	325954	ptsN_2
							325956	326393	dut
311682	310660	310559	311653	BB0302	892795	893807	892826	893983	ftsW
52610	51588	51253	52434	BB0056	794746	795789	794941	796152	pgk
28035	27140	27434	27865	BB0029	997284	998154	997122	997640	CT847
							997656	998162	CT848
298371	297476	297466	298776	BB0288	765474	766365	765053	766381	yscN
297347	296580	296428	297051	BB0286	989003	989618	988877	989842	mesJ
		297038	297469	BB0287					
490626	489861	489733	490554	BB0471	540112	540904	540292	540933	CT465
504707	503941	503926	504285	BB0494	588442	589308	588369	588866	rs5
		504298	504795	BB0495			588881	589252	r118
662403	661637	661606	662529	BB0630	690639	691280	690426	691121	CT610
6274	5508	5251	6312	BB0005	658647	659408	658617	659657	trpS
473539	472836	472566	473408	BB0453	541799	542519	541534	542592	atoS
505859	505220	505104	505541	BB0497	587606	588244	587942	588376	r115
179587	178948	178917	179543	BB0177	995073	995719	995075	995413	rs1
695299	694660	694693	695523	BB0655	384856	385501	385149	385610	CT338
343043	342404	342335	343207	BB0334	791980	792623	791730	792695	dppD
238978	238468	238301	239128	BB0234	997811	998377	997656	998162	CT848
365443	364932	365115	365603	BB0355	830284	830887	830165	830689	CT718
347011	346500	346431	346841	BB0338	141786	142289	141972	142361	rs9
189826	189317	189299	189859	BB0190	982158	982667	982118	982699	infC
50050	49540	49341	50012	BB0053	686297	686809	686330	687019	ung
411491	411015	410787	411446	BB0399	903475	903960	903584	903943	ybeB
690563	690180	690151	690489	BB0650	385115	385495	385149	385610	CT338
500995	500613	500593	501009	BB0485	592026	592407	592013	592429	r116
505027	504708	504799	505104	BB0496	589905	590193	589853	590254	rs8
438659	438404	438446	438557	5S_rrlB	858780	859118	858982	859098	5SrRNA_1
42883	42628	42480	42881	BB0044	898943	899200	898940	899272	rsbV_2
189059	188804	188708	189055	BB0188	982917	983169	982923	983294	r120
482307	482180	482222	482308	tRNA-Ser-3	485243	485361	485247	485330	tRNASer_3

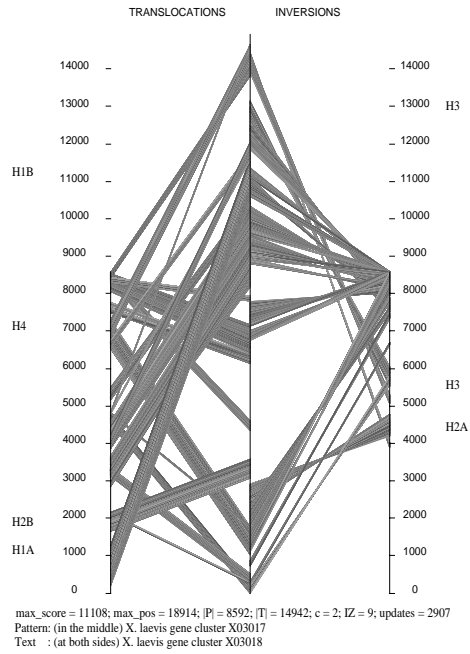


Figure 15: An alignment of two histone gene clusters from *Xenopus laevis*, GenBank accession number: X03017 (in the middle) and X03018 (at both sides). Note that genes H2A, H2B, H3, and H4 are marked on both sequences. The alignment shows that the orientation of H2A and H3 are reversed in the two sequences. This picture shows the Walking Tree Method is capable of finding inversions and translocations of genes.

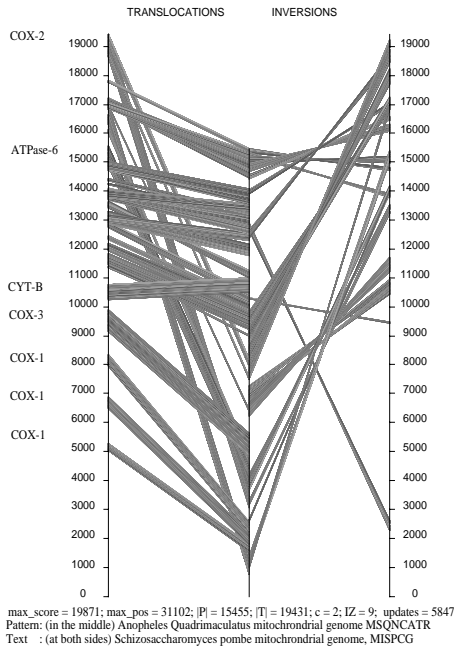


Figure 16: An alignment of the mitochondrial genomes of *Anopheles quadrimaculatus*, GenBank locus MSQNCATR (in the middle), and *Schizosaccharomyces pombe*, GenBank locus MISPCG (at both sides). The previously unrecognized Cytochrome c oxidase 3 (COX-3) region in this map is identified by the Walking Tree Method.

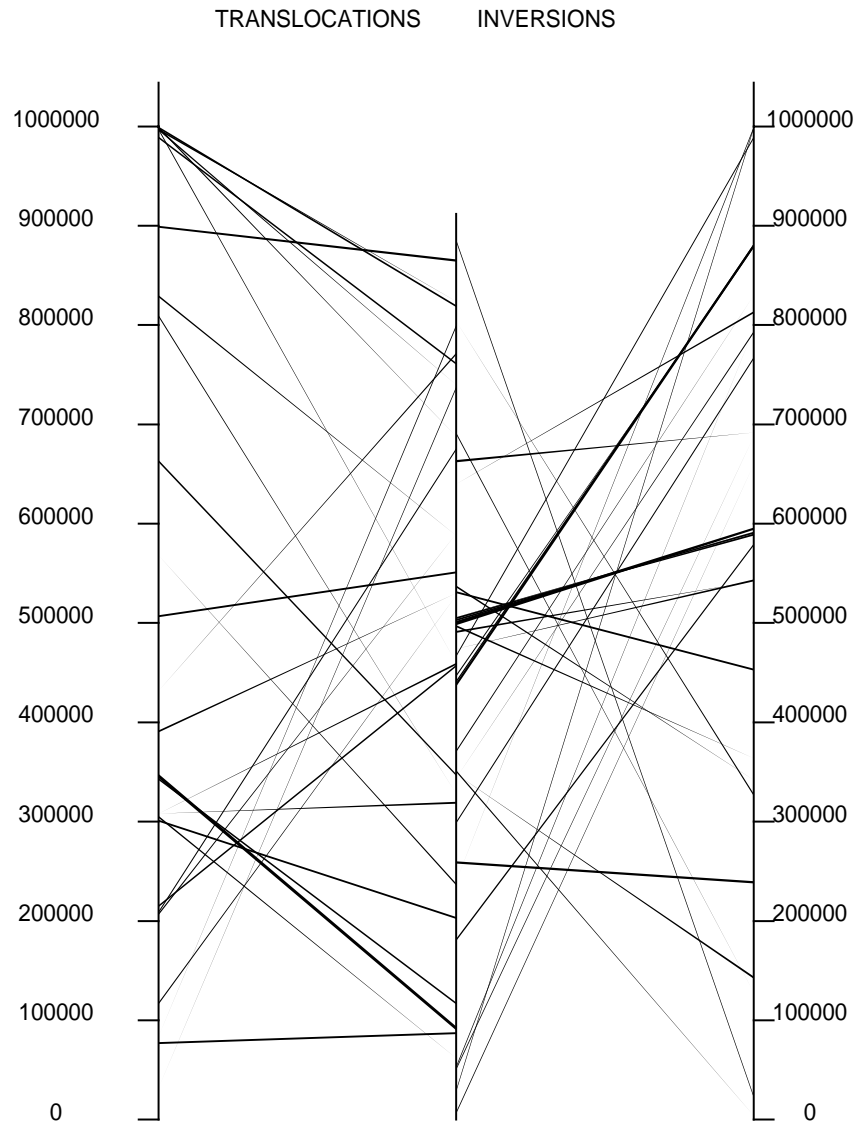


Fig 17. This picture shows that the Walking Tree Method reveals the matched genes that are labeled (annotated) on both DNAs (total DNA sequence of *Borrelia burgdorferi* aligned with the total DNA sequence of *Chlamydia trachomatis*). There are 40 translocations and 63 inversions in this picture. Again, this picture shows the Walking Tree Method is capable of finding inversions and translocations of genes.

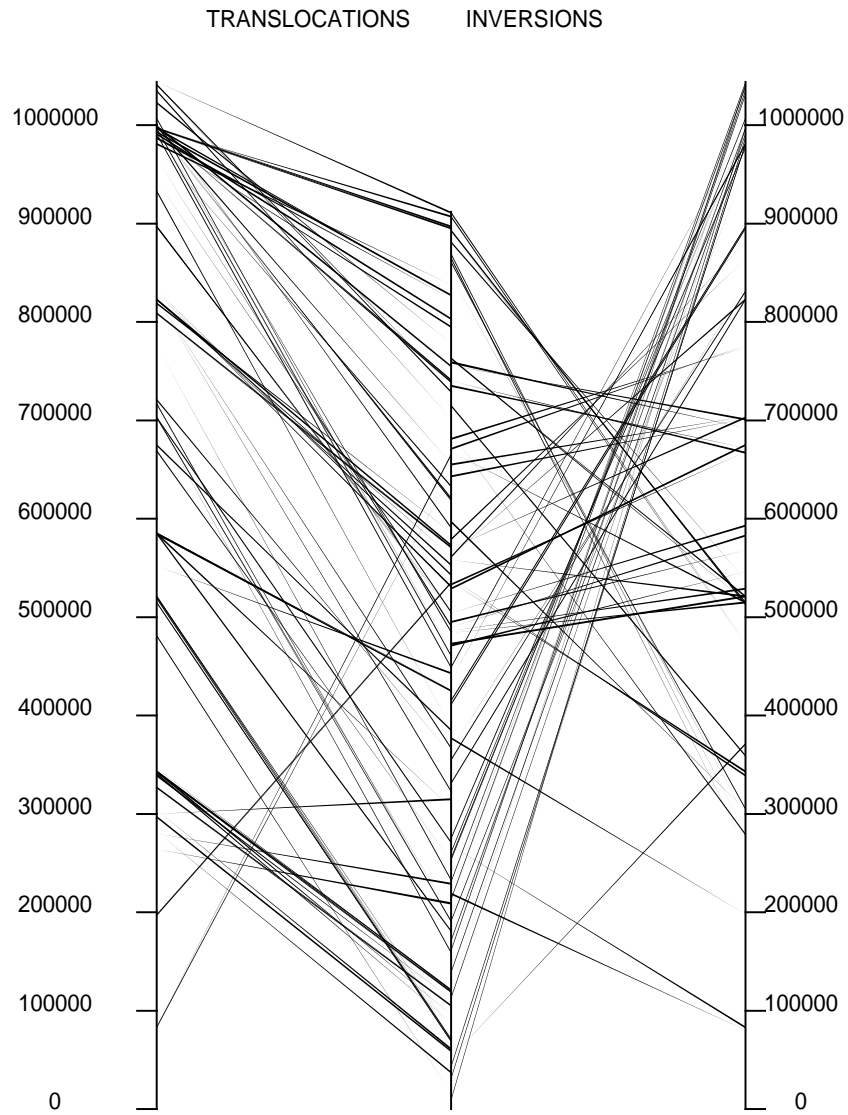


Fig 18. This picture shows that Walking Tree Method reveals the matched genes that are labeled on only one DNA, i.e., genes can be located in one sequence if the aligned portion of the other sequence is known to be a gene. There are 86 translocations and 62 inversions in this picture. This picture shows potential gene locations that are not annotated in one DNA, but annotated in another.

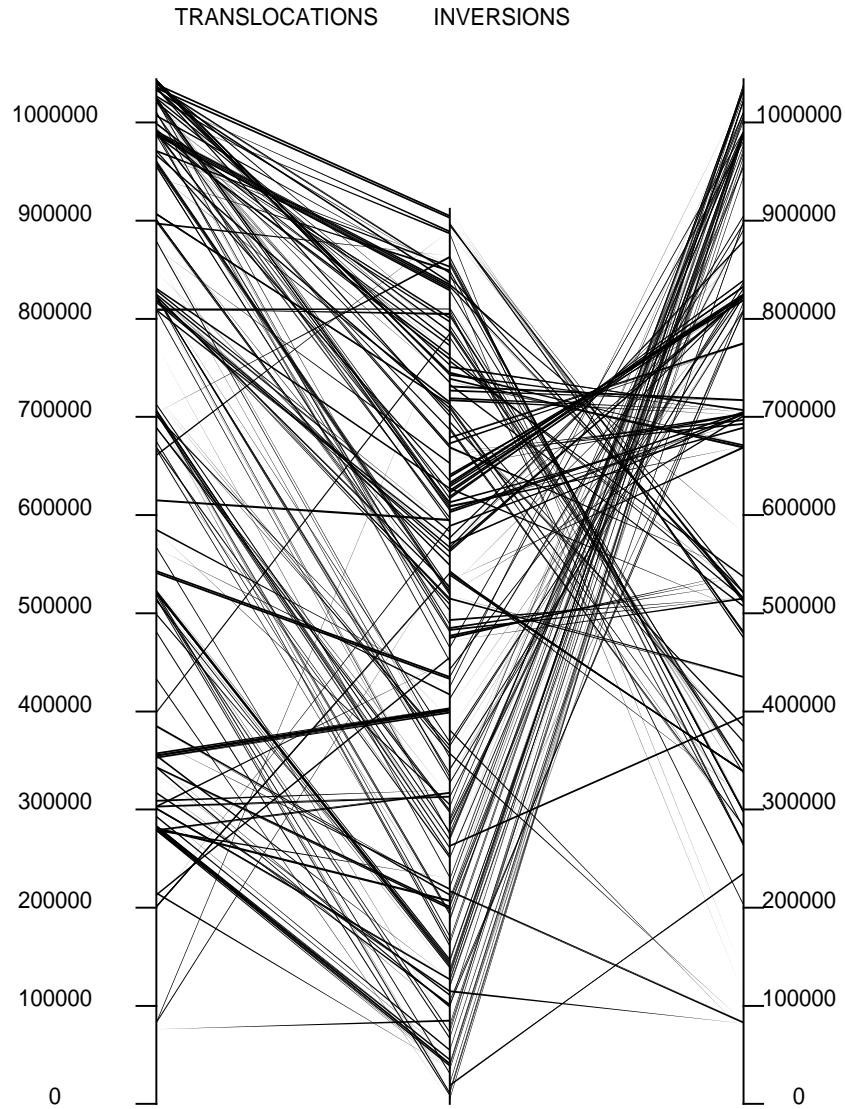


Fig 19. This picture shows that the Walking Tree Method reveals potential genes that are unlabeled on both DNAs. There are 700 translocations and 667 inversions in this picture. What interests us is the big match (*Chlamidia*: 352764 to 357294 and *Borrelia*: 399872 to 403967) which only covers 50% of the locus BORRPOB annotated in the GenBank database, but is found on both DNAs. This implies that *Borrelia*'s BORRPOB annotation in Genbank may need to be reinvestigated.

5. Conclusion

The Walking Tree Method is a powerful tool for gene finding. The technique works by finding a “best” alignment between sequences. In common with other techniques, the Walking Tree can use a known gene in one genome to find a corresponding gene in another genome.

The real power of the technique is to find corresponding but unannotated regions in different genomes. Preservation of regions across separated species is strong evidence of biological function. We gave several examples of the locations of genes or interesting regions in a variety of organisms. Our improved parallelization technique makes alignment of million base sequences a one day operation.

6. Reference

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