

1 Wheat Panache - a pangenome graph database representing
2 presence/absence variation across 16 bread wheat genomes

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17

18 **Abstract**

19 Bread wheat is one of humanity's most important staple crops, characterized by a large and complex
20 genome with a high level of gene presence/absence variation between cultivars, hampering genomic
21 approaches for crop improvement. With the growing global population and the increasing impact of
22 climate change on crop yield, there is an urgent need to apply genomic approaches to accelerate
23 wheat breeding. With recent advances in DNA sequencing technology, a growing number of high-
24 quality reference genomes are becoming available, reflecting the genetic content of a diverse range
25 of cultivars. However, information on the presence or absence of genomic regions has been hard to
26 visualize and interrogate due to the size of these genomes and the lack of suitable bioinformatics
27 tools. To address this limitation, we have produced a wheat pangenome graph maintained within an
28 online database to facilitate interrogation and comparison of wheat cultivar genomes. The database
29 allows users to visualize regions of the pangenome to assess presence/absence variation between
30 bread wheat genomes.

31 Database URL: http://www.appliedbioinformatics.com.au/wheat_panache

32 **Introduction**

33 Bread wheat (*Triticum aestivum*) is one of the most widely grown crops, yet there is a significant
34 challenge to increase yield to meet the projected demands of a growing world population. With
35 predictions of climate change-related yield losses ranging from 17% to 31% by the middle of the 21st
36 century (1), improved genomics-based breeding approaches are required to produce climate
37 change-ready wheat cultivars.

38 Wheat genomics has made rapid advances in recent years with the first draft genome assembly
39 produced in 2014 (2) based on the shotgun sequencing of isolated chromosome arms (3-5). A first
40 near-complete assembly of the variety Chinese Spring was produced in 2017 (6) with a final
41 reference genome assembly available in 2018 (7). This reference assembly was rapidly followed by
42 assemblies of fifteen additional cultivars from global breeding programs (8).

43 The increasing availability of reference genome assemblies made it clear that there is significant
44 presence/absence variation (PAV) between individuals (9-12). This insight has led to the production
45 of pangenomes that reflect the gene content of a species rather than an individual (10,13-20).
46 Pangenomes are now available for several plant species; the first bread what pangenome
47 representing the gene content of 16 bread wheat cultivars was published in 2017 (18). This wheat
48 pangenome was assembled using an iterative mapping approach, which efficiently identified new
49 gene space and called gene presence/absence between individuals. This kind of pangenome is
50 however limited in that the physical location of the new gene space can be difficult to determine
51 with accuracy. With the availability of multiple whole-genome references, this limitation may be
52 addressed through the production of a graph-based pangenome. Graph-based pangenomes have
53 recently become popular thanks to the graph data structure which can accurately represent the
54 physical locations of genomic and structural variants with minimal reference bias, with tools such as
55 vg (21), seqwish (22), minigraph (23), and PHG (24) being successfully applied to build variation,
56 sequence, or haplotype graphs.

57 A major limitation of pangenome graphs is that few tools are available to visualize these complex
58 graph structures. Genome visualization tools such as GBrowse (25), JBrowse2 (26), or Circos (27) are
59 designed to display information relative to a linear reference genome, not a graph of several
60 genomes, while graph viewers such as Bandage (28) or pangenome viewers such as ODGI (29) focus
61 on visualizing the graph itself, but display little other information such as genome annotations.

62 Panache is a recent pangenome visualization tool that can process linearized assembly graphs and
63 display shared regions as a web-based dynamic heatmap (30). Panache has so far only been applied
64 to visualize presence/absence variation in the banana pangenome (14), but has the potential to be
65 expanded to other species, even for crop genomes as large as wheat. Here, we present a graph
66 pangenome representing 16 bread wheat cultivars hosted within a public Wheat Panache database,
67 with a new web-based browser for visualizing genomic regions across the wheat pangenome, along

68 with the graph formatted for minimap2 (31) and Giraffe (32). This tool offers researchers and
69 breeders the ability to assess genome variation between these varieties, mining the diversity present
70 in this large and complex genome.

71 Results and discussion

72 A wheat graph pangenome

73 We constructed a graph pangenome using 16 high-quality wheat genome assemblies Representing
74 the global variation of modern bread wheat cultivars. The assembled graph had a total size of 15.8
75 Gbp in comparison with the founder genome assembly sizes of 13.9 to 14.2 Gbp (33). After aligning
76 all genomes back to the graph, these 15.8 Gbp were split up into 2,791,482 segments present in at
77 least one individual. The segments had an average size of 5.6 Mbp (median: 498 bp), ranging from 2
78 bp to 37.6 Mbp (Figure 1A). Realignment of the 16 genome assemblies to the graph revealed that
79 out of the 2.7 million segments, 542,711 (19%) segments were present in all individuals (total size:
80 10.2 Gbp ranging from 2 bp to 4.9 Mbp) with the remaining 2,248,771 segments (total size: 5.6 Gbp)
81 being present in a median of 8 individuals (Supplementary Figure 1). 10,437 segments (0.4% of all
82 segments) with a total length of 19.9 Mbp (average length: 1.9 Kbp) were not covered by any
83 genome assembly during the realignment step, probably due to these segments being too small
84 and/or too repetitive.

85 Interestingly, the cultivar with the most unique segments was the reference cultivar Chinese Spring,
86 with 158,503 (7%) of segments with a total size of 140.5 Mbp being only present in Chinese Spring
87 (Figure 1B). This may be due to the genomic distance between Chinese Spring and the other cultivars,
88 consistent with previous observations (18), and reflecting Chinese Spring's age (collected around
89 1900) and its lack of agronomic characters that were selected for in modern cultivars (34). The
90 distance between the Chinese Spring assembly and the 15 other assemblies is also supported by 1.2
91 Gbp of the graph in 901,475 segments not being present in Chinese Spring but in at least one other
92 cultivar, reflecting the complex history of introgressions in modern bread wheat (8,35). We aligned

93 the IWGSC v1 gene annotation for Chinese Spring (7) back to the graph by intersecting the linearized
94 graph with gene positions. We found a position in the graph for 110,790 (100%) genes confirming
95 that the graph assembly contains all gene models of the IWGSC assembly.

96 **The Wheat Panache web portal**

97 Using this graph, we built a web-based Panache instance (30), allowing users to visualize regions or
98 genes of interest for presence/absence across the chosen wheat cultivars. The webserver is available
99 at http://www.appliedbioinformatics.com.au/wheat_panache.

100 Wheat Panache displays a linear version of the pangenome graph subdivided into blocks based on
101 presence/absence of the selected individuals. A block is defined to have no internal
102 presence/absence variation and to contain at least one gene. Blocks are named based on the
103 pseudomolecule they originated in, and as we started the assembly with the IWGSC assembly, most
104 blocks (1,890,035 out of 2,791,483 blocks, 67%) are named after their position in the IWGSC
105 assembly.

106 The interface displays the linearized pangenome as a chain of such graph segments, with one
107 horizontal track per cultivar (Figure 2). Coordinates are based on the pangenome graph assembly.
108 Genes are represented as black dots above blocks and hovering over a gene reveals its coordinates
109 within the assembly and exon structure. Three summary tracks below the cultivar tracks show which
110 blocks are core or variable based on a user-definable threshold, how long the block is, and how often
111 the block is repeated within Panache. Users can zoom into blocks, or search for ‘hollow areas’ (areas
112 of consecutive absence based on a user-defined threshold) using the Hollow Area Finder, which is a
113 convenient way to automatically focus on large PAV areas. Users can sort the cultivars
114 alphanumerically, by gene presence/absence status, or by a phylogeny based on Mash v2.3 (36). The
115 graph assembly displayed in Wheat Panache, including a version pre-indexed for vg v1.37.0’s Giraffe
116 (32) is available at <https://doi.org/10.5281/zenodo.6085239> (37) allowing for downstream analyses
117 of the population graph.

118 In summary, we present the first wheat graph pangenome assembly, based on 16 cultivars with an
119 online visual representation of the graph within the Panache visualization tool. The graph assembly
120 will be a valuable tool for wheat genomics researchers looking for a more accurate reference
121 assembly. The web platform Panache allows users to interrogate this graph and search for structural
122 variants around regions of interest.

123

124 **Materials and Methods**

125 We used publicly available genome assemblies, including fifteen high-quality *Triticum aestivum*
126 genome assemblies (8) and the IWGSC v1 *T. aestivum* cv. Chinese Spring assembly (7), to assemble a
127 graph using minigraph v0.14 (23). To optimize assembly, we used k-mers that appear fewer than 100

128 times (-f.1) for the graph assembly and assembled the graph genome by genome, starting with

129 IWGSC v1 followed by alphabetical order of cultivar names, ending with the *T. aestivum* ssp. *spelta*

130 PI190962 assembly.

131 All assemblies were aligned with the final graph using minimap2 v2.18 (31) and alignments were

132 converted to BED format. The main graph was linearized using gfatools gfa2bed v0.4 with default

133 parameters (<https://github.com/lh3/gfatools/releases>) and merged with all minimap2 alignments

134 using bedtools v2.30.0 multiinter (38). The resulting blocks were intersected with the IWGSC gene

135 annotation using bedtools v2.30.0 intersect.

136 The data was converted to Panache JSON format and a Panache instance was set up to serve the

137 data (30). To make the display feasible on a regular workstation, we retained only blocks overlapping

138 with Chinese Spring genes and then merged adjacent blocks if they showed identical PAV behaviour

139 across all individuals.

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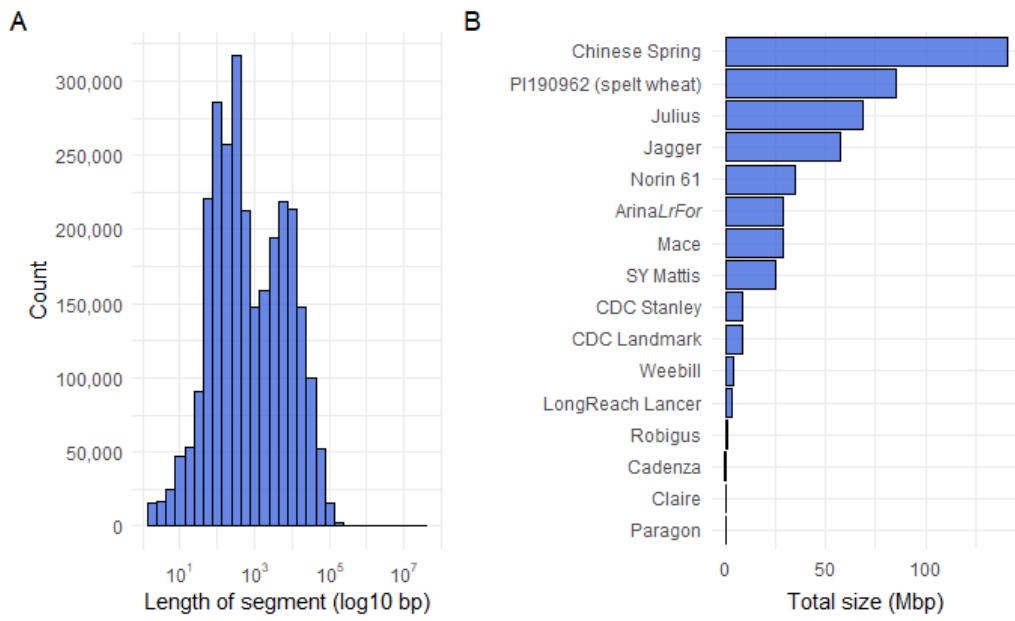
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234 **Figures**



235

236 Figure 1: A) Bar chart showing the distribution of the size of all assembly graph segments (log-scale).

237 B) Total size of unique segments per cultivar in Mbp. PI190962 is a line of species *Triticum spelta*,

238 Chinese Spring is the reference cultivar of *T. aestivum*.

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242 Figure 2 – Wheat Panache screenshot showing a *Aegilops ventricosa* introgression at the beginning
243 of chromosome 2 in cultivars Stanley, Jagger, Mace, and SY Mattis (39,40). Black boxes were added
244 to show the region missing in cultivars where the introgression replaced parts of 2A. The graph
245 assembly started with the IWGSC v1 assembly leading to linearized regions following the same
246 naming scheme as the IWGSC v1.0 assembly (chr1A_part1, chr1A_part2, chr2A_part1, ...). CS stands
247 for Chinese Spring. Shown here is the beginning of the first part of chr2A. Black blocks are gene
248 models. White regions correspond to regions that are present in the graph but contain no genes.

249