GenAPI: a tool for gene absence-presence identification in fragmented bacterial genome sequences

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Abstract

Background: Bacterial gene loss and acquisition is a well-known phenomenon which contributes to bacterial adaptation through changes in important phenotypes such as virulence, antibiotic resistance and metabolic capability. While advances in DNA sequencing have accelerated our ability to generate short genome sequence reads to disentangle phenotypic changes caused by gene loss and acquisition, the short genome sequence reads often results in fragmented genome assemblies as a basis for identification of gene loss and acquisition events. However, sensitive and precise determination of gene content change for fragmented genome assemblies remain challenging as analysis needs to account for cases when only a fragment of the gene is assembled or when the gene assembly is split in more than one contig.

Results: We have developed GenAPI, a command-line tool that is designed to compare the gene content of closely related bacterial genomes for which only fragmented genome assemblies are available. GenAPI, unlike other available tools of similar purpose, accounts for genome assembly imperfections

and aims to compensate for them. We tested the performance of GenAPI on three different datasets to show that GenAPI has a high sensitivity while it maintains precision when dealing with partly assembled genes in both simulated and real datasets. Furthermore, we compared and evaluated the performance of GenAPI with six popular tools for gene presence-absence identification. While we find that the compared tools have the same precision and recall rates when analyzing complete genome sequences, GenAPI performs significantly better than the other tools on fragmented genome assemblies.

Conclusions: We have developed a bioinformatics tool, GenAPI, to identify and compare the gene content of bacteria for which only fragmented genome assemblies is available, and we showed that GenAPI identifies gene presence-absence in fragmented genome assemblies with higher precision in comparison to other tools of similar purpose while not compromising the sensitivity of the analysis.

Keywords: sequence analysis, bacterial genomics, evolution

Background

Finding differences in the gene repertoire between bacterial clones is important to understand the genetic basis of differences in phenotypes such as virulence, antibiotic resistance, and metabolic capability. Also, phylogenetic analysis including information about the absence or presence of genes helps to inform about the pace and mechanisms by which genes are lost and acquired in bacterial populations [1]. Thus, genome-wide analysis of the gene presence or absence is necessary for a better understanding of bacterial evolution and adaptation.

There exist multiple open-source bioinformatics tools available for gene presence-absence identification. Each tool has its own set of advantages and limitations. Here we focus on tools that use assembled genomes as input. Some tools, e.g. PanSeq (Laing, et al., 2010), are based on alignment of the query genome sequence to a reference genome to specifically test for the presence-absence of genes that are present in the given reference. Accordingly, this approach is limited knowing that prophage and plasmid genes constitute a major part of the variable bacterial genetic content [2]. Other tools, e.g. Roary [3], SaturnV [4], PanDelos [5], panX [6], BPGA [7] and EDGAR [8] construct a pan-genome from inputted genome assemblies and then determine the gene set that is present in each of the genome assemblies. The performance of both approaches depends on the completeness of the queried genome sequences, and all previously mentioned tools are designed for analysis of near-complete genomes with only minor parts of genes missing due to sequencing and assembly imperfections. As a result, there is a need for analytical tools that are designed to account for highly fragmented assemblies (often the product of *de novo* genome assemblies based on short-read sequence data) that else would result in a large number of false positive calls for gene being absent in the assembly.

Here, we introduce GenAPI, a command line tool for identification and comparison of the gene content in bacterial clones of the same species. We specifically developed GenAPI to work successfully on fragmented genome assemblies to account for cases when only a fragment of the gene is assembled or when the gene assembly is split in more than one contig. The compensation for imperfections of sequencing and assembly proves to minimize the false positive gene absence calls while maintaining the true positive gene absence call rates.

Results

Description of GenAPI algorithm

GenAPI is designed to run in a Unix environment and requires BLAST+, CD-HIT and Bedtools softwares [9, 10, 11]. R with pheatmap library [12, 13] and RAxML [14] software is needed for optional gene presence-absence matrix visualization and phylogenetic tree generation. The step-wise workflow of GenAPI is shown in Figure 1A where annotated genome sequence files are taken as an input and the pan-genome is generated by clustering the genes with CD-HIT-EST with 90% identity and 80% sequence length overlap requirements. The best alignments between genes of the pan-genome and each genome sequence are identified. Finally, the genes are defined as present or absent according to the user provided thresholds (default: gene is present if best alignments have minimum 25% coverage with 98% identity or 50% coverage with 90%). More detailed method explanation is provided in Supp. Figure 1.

The requirement of a gene to be only partly aligned, but with high identity, to be defined as present allows to compensate for sequencing and assembly imperfections while not compromising precision. Nonetheless, genes shorter than 150 base pairs (bp) are recommended to be excluded from the gene presence-absence analysis as alignment of short gene sequences may produce unspecific alignments that

Figure 1. GenAPI workflow. GenAPI data flow including the tools [10]ᵃ, [9]ᵇ or algorithms used in each step (a), output gene presence-absence matrix (b) and phylogenetic tree (c).

may pass threshold for defining a gene as present (e.g. a 100 bp gene would be defined as present with

an unspecific 25 bp alignment with 98% identity). Short genes are excluded from the analysis by default.

Furthermore, GenAPI is developed for identification of gene sequences that are completely lost or acquired in the genome, therefore it will not identify copy number changes of multicopy genes (including pseudogenes) or partial gene deletions.

GenAPI outputs several files: (1) a pan-genome file with all gene sequences, (2) a gene presence-absence matrix with information on the presence of each gene in each of analyzed genome sequences, and (3) a list of best alignment statistics for each gene in each analyzed genome sequence. Furthermore, a heatmap visualization of gene presence-absence across all genomes and a maximum-likelihood phylogenetic tree based on gene presence-absence information is implemented in GenAPI (Figure 1B-C).

Evaluation of GenAPI and comparison to other tools for gene presence-absence identification

We evaluated the performance of GenAPI on three test datasets with genome assemblies; two simulated datasets with *in silico* introduced variation in gene content, and one real dataset with known gene deletions. Furthermore, we compared the performance of GenAPI against Roary, SaturnV, PanDelos, BPGA, panX and EDGAR which are tools developed for similar, but not identical, purposes. None of the tools which we compared against were developed for fragmented genome assemblies, however, (1) fragmented genome assemblies are most often the output from genome sequencing experiments, (2) there are no available tools developed for gene presence-absence analysis in fragmented genome assemblies.

The results of the predicted gene absence of the recall, precision and F1 scores are shown in Table 1. F1 measure was calculated in order to summarize precision and recall scores as it is a harmonic average of these two measures. Recall, precision and F1 scores were calculated by using the following formulas:

$$Recall = \frac{TP}{TP+FP}; Precision = \frac{TP}{TP+FN}; F1 = 2 \cdot \frac{Precision \cdot Recall}{Precision+Recall}.$$

For simulated *P. aeruginosa* and long-term *E. coli* experiment [15] datasets, Prokka version 1.11 [16] was used for gene annotation. Sequencing reads from both datasets were assembled using SPAdes version 3.10.1 software with default parameters. [17] *P. aeruginosa* and *E. coli* assembly statistics are reported in Supp. Table 1 and Supp. Table 2 respectively. Information about known deletions in *E. coli* experiment dataset was obtained from Barrick *et al*., 2009 [15] study. *S. typhi* dataset was preprocessed as described by Page *et al.*, 2015 [3]. All tools were tested with default parameters, with an exception for Roary for which paralog splitting was disabled since the other tools do not split identical sequences (paralogs). Default parameters were chosen assuming that they are the most optimal parameters for the tool. Sequencing reads for *P. aeruginosa* simulated dataset were simulated using ART software [18] with 150bp paired end sequences with 200bp insert sizes and 20x average genome coverage.

Table 1. Recall, precision and F1 score of GenAPI, panX, BPGA, Roary, PanDelos, EDGAR and SaturnV using simulated and real datasets.

		GenAPI	panX	BPGA	Roary	PanDelos	EDGAR	SaturnV
S. typhi [a]	Precision	1	NA	0.93	1	NA	1	1
	Recall	1	NA	1	1	NA	1	1
	F1	1	NA	0.97	1	NA	1	1
P. aeruginosa [b]	Precision	0.94	0.37	0.31	0.31	0.45	0.15	0.11
	Recall	1	1	0.94	1	1	1	1
	F1	0.97	0.54	0.46	0.47	0.62	0.26	0.20
E. coli [c]	Precision	0.95	0.47	0.26	0.23	0.21(0.21)	0.12	0.09
	Recall	0.98	1	0.88	1	0.71(1)	1	1
	F1	0.97	0.64	0.40	0.38	0.32(0.35)	0.21	0.17

a - simulated *S. typhi* dataset with complete genome sequences used in Roary publication [3]; b - simulated *P. aeruginosa* dataset with partly assembled gene instances; c – real *E. coli* dataset [15].

First, all tools except panX and PanDelos were tested on the same simulated *Salmonella typhi* dataset that was used for the evaluation of Roary in its own publication by Page et al., (2015) [3]. PanDelos and

panX were excluded from the analysis as the available input files were not compatible with the input requirements of these two tools. All five included tools identified all 181 instances where genes were absent from genomes (GenAPI did not include one gene as it was shorter than the default 150bp gene length requirement) (Supp. Table 3). BPGA additionally made 12 false calls of gene being absent. The other tools did not falsely call any genes to be absent.

Second, we simulated sequencing reads for a dataset of 8 *Pseudomonas aeruginosa* genome sequences with known deletions in order to test the tools on data representing draft genome assemblies (available at https://github.com/MigleSur/GenAPI). In total there were 49 absent genes (Supp. Table 5). All tools except BPGA, which missed 3 deletions, correctly identified all 49 absent genes. However, false positive calls were made by all tools. 145, 387 and 103 genes were false positively called as absent by Roary, SaturnV and BPGA, respectively. PanDelos and panX performance was better and resulted in 60 and 83 genes to be false positively called as being absent, respectively. On the other hand, only 3 false positive calls were made by GenAPI (Supp. Table 3).

Finally, we tested the tools on a dataset of 6 genome sequences with 102 known deletions from the *E. coli* long-term evolution experiment performed by Richard Lenski [15]. Roary, SaturnV, EDGAR and PanX called all true deletions, and GenAPI closely followed these tools except missing two copy number changes of homologous prophage genes. BPGA and PanDelos failed to identify 12 and 30 true deletions, respectively. The number of false positive calls of gene deletions was observed to be high for Roary, SaturnV, panX, PanDelos, BPGA and EDGAR (115-1009 false positive deletion calls) in contrast to GenAPI (5 false positive calls) (Supp. Table 3). F1 scores from Table 1 show a summary of how well each tool performs with each of the datasets.

Overall, we find that the compared tools have the same precision and recall rates when analyzing complete genome sequences (*S. typhi* dataset); whereas GenAPI performs significantly better than the other tools on fragmented genome assemblies in *P. aeruginosa* and *E. coli* datasets (Table 1, Supp. Table 3).

Conclusions

We have developed a bioinformatics tool, GenAPI, to identify and compare the gene content of bacterial genomes of the same species. While there exist other tools with similar purposes, GenAPI is specifically designed for the analysis of fragmented, closely related genome sequences. Fragmented genome sequences are often the product of *de novo* genome assembly based on short-read sequence data, and GenAPI will therefore be suitable for use in a large number of studies including bacterial genome sequence data. We tested the performance of GenAPI on three different datasets to show that GenAPI has a high sensitivity while it maintains precision when dealing with partly assembled genes in both simulated and real datasets. We furthermore have shown that none of the other popular gene presence-absence identification tools are well suited for analysis of fragmented genome assemblies and GenAPI's performance is better with fragmented genomes.

Availability and requirements

Project name: GenAPI

Project home page: https://github.com/MigleSur/GenAPI

Operating system(s): e.g. Platform independent

Programming language: UNIX

Other requirements: BLAST 2.6.0+ or higher, CD-HIT 4.6.1 or higher, Bedtools 2.26 or higher

License: GPLv3

Any restrictions to use by non-academics: None

Declarations

Ethics approval and consent to participate: Not applicable

Consent for publication: Not applicable

Availability of data and material: The datasets generated and analysed during the current study are available in the test_samples directory in GenAPI repository, https://github.com/MigleSur/GenAPI/ ; The datasets used and analyzed during the current study are available from the corresponding author on reasonable request.

Competing interests: The authors declare that they have no competing interests

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Authors' contributions: MG developed the tool, MG and RLM analyzed and interpreted the results. MG and RLM were involved in writing the draft, reading and approving the final manuscript.

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