INTEGRATE-Vis: a tool for comprehensive gene fusion visualization

Jin Zhang
Washington University School of Medicine in St. Louis

Teng Gao
Washington University School of Medicine in St. Louis

Christopher A. Maher
Washington University School of Medicine in St. Louis

Follow this and additional works at: https://digitalcommons.wustl.edu/open_access_pubs

Recommended Citation
Zhang, Jin; Gao, Teng; and Maher, Christopher A., "INTEGRATE-Vis: a tool for comprehensive gene fusion visualization." Scientific Reports.7, (2017).
https://digitalcommons.wustl.edu/open_access_pubs/6426
INTEGRATE-Vis: a tool for comprehensive gene fusion visualization

Jin Zhang¹, Teng Gao²,⁵ & Christopher A. Maher²,³,⁴,⁶

Despite the increasing quantity of tools for accurately predicting gene fusion candidates from sequencing data, we are still faced with the critical challenge of visualizing the corresponding gene fusion products to infer their biological consequence (i.e. novel protein and increased gene expression). This is currently accomplished by manually inspecting and inferring the biological consequence of top scoring gene fusion candidates. This labor-intensive process could be made easier by automating the annotation of gene fusion products and generating easily interpretable visualizations. We developed a gene fusion visualization tool, called INTEGRATE-Vis, that generates comprehensive, highly customizable, publication-quality graphics focused on annotating each gene fusion at the transcript- and protein-level and assessing expression within an individual sample or across a patient cohort. INTEGRATE-Vis is the first comprehensive gene fusion visualization tool to help a user infer the potential consequence of a gene fusion event. It has potential utility in both research and clinical settings. INTEGRATE-Vis is available at https://github.com/ChrisMaherLab/INTEGRATE-Vis.

Gene fusions have served as highly specific diagnostic markers, prognostic indicators and therapeutic targets¹. High throughput transcriptome sequencing (RNA-Seq) has accelerated our ability to discover expressed gene fusions². While recent tools, such as INTEGRATE³, are highly sensitive and specific, we are still faced with the critical challenge of ensuring that a casual gene fusion is not only detected, but that it can be prioritized accordingly amongst passenger events. This is currently accomplished by manually inspecting and inferring the biological consequence (i.e., generation of a novel protein, altered expression levels) of top scoring gene fusion candidates. The labor-intensive process could be made easier and more precise by automating the annotation of gene fusion transcripts and proteins and generating easily interpretable visualizations. Currently gene fusion visualization approaches either rely on CIRCOS to highlight the genomic locations of the gene fusion partners⁴, IGV for assessing sequence coverage at fusion junctions⁵, or splicing graphs to observe the exons involved in a gene fusion⁶. Individually each of these methods is insufficient for inferring the consequences of the gene fusion on expression or the corresponding protein product. To address these limitations, we developed a tool, INTEGRATE-Vis, which generates multiple visualizations for annotating each gene fusion, at the transcript- and protein-level, and assessing gene expression within an individual sample or across a cohort.

Results

INTEGRATE-Vis pipeline (Fig. 1) generates four types of figures, created manually in previous publications, to provide easy-to-interpret visualizations of gene fusion predictions⁷–¹⁰. To illustrate how INTEGRATE-Vis works we focused on a prostate cancer patient harboring the most prevalent gene fusion, TMPRSS2-ERG, that results in the marked increase in the expression of the oncogenic transcription factor ERG (Fig. 2). Panels A through D of Fig. 2 correspond to structure plot, domain plot, exon expression plot, and gene expression plot outputs described in Fig. 1, respectively.

¹Department of Radiation Oncology, Washington University School of Medicine, St. Louis, Missouri, 63110, USA. ²McDonnell Genome Institute, Washington University School of Medicine, St. Louis, Missouri, 63110, USA. ³Department of Internal Medicine, Division of Oncology, Washington University School of Medicine, St. Louis, Missouri, 63110, USA. ⁴Siteman Cancer Center, Washington University School of Medicine, St. Louis, Missouri, 63110, USA. ⁵Department of Computer Science and Engineering, Washington University, St. Louis, Missouri, 63105, USA. ⁶Department of Biomedical Engineering, Washington University, St. Louis, Missouri, 63105, USA. Correspondence and requests for materials should be addressed to C.A.M. (email: christophermaher@wustl.edu)
Figure 1. Overview of the INTEGRATE-Vis pipeline. INTEGRATE-Vis contains four major modules to plot four types of figures for gene fusions: isoform structure, domain, exon expression, and gene partner expression. The first three modules are for individual samples, and the last module is for cohort data. Only minimal inputs in standard formats, including BEDPE, TSV, FASTA, BAM, and GTF are needed to make the plots.

Figure 2. INTEGRATE-Vis output illustrated using the TMPRSS2-ERG gene fusion in prostate cancer. INTEGRATE-Vis outputs four visualizations including: (A) gene fusion transcript isoforms, (B) the predicted protein structure of the gene fusion, (C) RNA-Seq read coverage across each gene fusion partner to reveal changes in exon expression (A red line is plotted at the fusion junctions at both gene partners. Exon boundaries are represented by blue lines. A marked expression change occurs between exons 3 and 4 of ERG.), and (D) expression of each gene fusion partner across the TCGA PRAD cohort. Blue is used to represent supporting reads, exons, transcript, and genomic locations for the 5' gene partner (TMPRSS2), while red is for those of the 3' gene partner (ERG).
First, in our structure plot we show the predicted gene fusion transcript structure highlighting sequence reads that encompass and span the fusion junction (Fig. 2A). Both TMPRSS2 and ERG are on the reverse strand of chromosome 22 in two consecutive cytogenetic bands. A genomic deletion between the upstream gene, TMPRSS2, and the downstream partner, ERG, generates the gene fusion event. By default, INTEGRATE-Vis constructs a gene fusion transcript isoform using the most prevalent transcript isoform of the gene partners. Alternatively, a user can designate specific transcript isoforms to display in the reconstructed gene fusion transcript. As shown in Fig. 2A, a 14-exon isoform (ENST00000458356) of TMPRSS2 and a 12-exon isoform (ENST00000398919) of ERG were used for visualization. The fusion junction is located at the second exon of ENST00000458356 and the fourth exon of ENST00000398919. The gene fusion transcript is shown in the upper panel with the corresponding supporting reads to infer the expression level. To conserve space, INTEGRATE-Vis displays 1, 2, or 3 supporting reads to illustrate 1, 2–10, or >10 supporting sequence reads, respectively.

Second, to predict the potential functional consequences of a gene fusion INTEGRATE-Vis generates a domain plot to translate the fusion transcript and displays the corresponding protein domains (Fig. 2B). The protein product from the 5′ gene partner (i.e. TMPRSS2) is plotted on the left and the protein product from the 3′ gene partner (i.e. ERG) is plotted on the right. This is also indicated by the annotation of N and C for N- and C- terminuses (Fig. 2B). As shown in Fig. 2B, the in-frame gene fusion protein product is comprised of a small 5′ regulatory region of TMPRSS2 and the majority of ERG. This includes both the ETS domain and Pointed domain of the ERG gene. As shown in Figure S1, for out-of-frame gene fusion transcripts, the 3′ end is represented by a white box and a red cross is used to represent the translation termination site.

Third, to determine whether the gene fusion increases the expression of the 3′ partner (i.e., oncogene) or decreases the expression of the 5′ partner (i.e., tumor suppressor), INTEGRATE-Vis generates an exon expression plot. This displays read coverage for the exon-level read coverage compared to the exons that are not included in the gene fusion (exons 1 through 3). Notably, INTEGRATE-Vis automatically selects the scales for the y- and x-axes to show the ranges of read coverage for each gene partner, although they can be adjusted based on user-defined input (Figure S2).

Fourth, to determine if the sample harboring the gene fusion results in a unique expression change relative to a cohort of samples (e.g., prostate cancer patients lacking the gene fusion), INTEGRATE-Vis generates a gene expression plot. This outputs a bar plot of the expression level for both genes involved in the fusion across a patient cohort (Fig. 2D). For example, Fig. 2D highlights the difference in ERG expression levels in patients lacking the TMPRSS2-ERG gene fusion with patients harboring the TMPRSS2-ERG gene fusion. In contrast, expression levels of TMPRSS2 are not different between patients with or without the gene fusion. In addition to determining if the gene fusion alters the expression of the 5′ or 3′ gene, this visualization can also identify additional patients that may also harbor gene fusions producing similar expression consequences.

While we have demonstrated the utility of INTEGRATE-Vis using the most prevalent TMPRSS2-ERG isoform, INTEGRATE-Vis automatically generates plots for all predicted gene fusion isoforms. INTEGRATE-Vis has been implemented with reasonable default parameters to help best interpret the functions of the gene fusion products. It also provided ample options to enhance user-friendliness (Figures S2 and S3). INTEGRATE-Vis executes efficiently; figure generation takes a few seconds (Figure S4).

Discussion
Overall, we developed the first comprehensive gene fusion visualization tool, INTEGRATE-Vis, which generates publication-quality graphics to help a user infer the potential consequence of a gene fusion event. We have implemented INTEGRATE-Vis to utilize standardized input files, including the SMC-RNA BEDPE format for gene fusion predictions, therefore making it widely accessible to the larger research community independent of the gene fusion discovery tool being used.

Methods
The INTEGRATE-Vis pipeline was implemented in Python and C++, and requires a minimal set of dependencies (CMake, GCC, Matplotlib, and gtfToGenePred) to install and execute. The input into INTEGRATE-Vis includes a list of gene fusion candidates in a standard BEDPE format as well as other common standardized inputs (i.e. FASTA, GTF) including a reference genome and gene models. 333 BEDPE files can be downloaded from https://github.com/ChrisMaherLab/INTEGRATE-Vis, including gene fusions previously discovered15. Additional input files in TSV format (i.e. a protein domain table and an ideogram table for cytogenetic bands) and the commands to generate these TSV files are all included at https://github.com/ChrisMaherLab/INTEGRATE-Vis. Read counts for the samples were calculated using FeatureCounts16. INTEGRATE-Vis performs a series of annotation and calculation steps before generating figures summarizing the gene fusion in PDF format (Fig. 1).

Availability and requirements. The INTEGRATE-Vis pipeline has been tested using Python version 2.7 and requires CMake, GCC, Matplotlib, and gtfToGenePred to install and execute. It is available from https://github.com/ChrisMaherLab/INTEGRATE-Vis, which also contains instructions and links for downloading required tools or packages, step-by-step instructions of installing INTEGRATE-Vis pipeline, and sample command lines of executing INTEGRATE-Vis from either BEDPE files or raw RNA-seq reads. Raw sequence reads of TCGA PRAD cohort can be downloaded from Genomic Data Commons (https://gdc.cancer.gov).
References


Acknowledgements

This work was supported by an NIH National Cancer Institute R21CA185983-01 (to C.A.M.), NIH National Cancer Institute R00CA149182 (to C.A.M.) and a Prostate Cancer Foundation Young Investigator Award.

Author Contributions

J.Z. and C.A.M. conceived the study. J.Z. and T.G. implemented the source code. J.Z. and C.A.M. wrote the manuscript and created the figures. All authors read and approved the final manuscript.

Additional Information

Supplementary information accompanies this paper at https://doi.org/10.1038/s41598-017-18257-2.

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

Publisher’s note: Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

Open Access This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made. The images or other third party material in this article are included in the article’s Creative Commons license, unless indicated otherwise in a credit line to the material. If material is not included in the article’s Creative Commons license and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this license, visit http://creativecommons.org/licenses/by/4.0/.

© The Author(s) 2017