CREPE: a Shiny app for transcription factor cataloguing

Diego A. Rosado-Tristani and José A. Rodríguez-Martínez *

Department of Biology, University of Puerto Rico Río Piedras, San Juan 00925, Puerto Rico

*To whom correspondence should be addressed.

Associate Editor: Alex Bateman

Received on February 13, 2023; revised on March 31, 2023; editorial decision on April 5, 2023; accepted on April 20, 2023

Abstract

Summary: Transcription factors (TFs) are proteins that directly interpret the genome to regulate gene expression and determine cellular phenotypes. TF identification is a common first step in unraveling gene regulatory networks. We present CREPE, an R Shiny app to catalogue and annotate TFs. CREPE was benchmarked against curated human TF datasets. Next, we use CREPE to explore the TF repertoires of Heliconius erato and Heliconius melpomene butterflies.

Availability and implementation: CREPE is available as a Shiny app package available at GitHub (github.com/diros-tri/CREPE).

Contact: jose.rodriguez233@upr.edu

Supplementary information: Supplementary data are available at Bioinformatics Advances online.

1 Introduction

Transcription factors (TFs) are sequence-specific DNA-binding proteins that regulate transcription. TF proteins bind to cis-regulatory elements (CREs), genomic regions that regulate the expression of nearby genes. Examples of CREs include promoters, enhancers and silencers (Wittkopp and Kalay, 2011). By regulating spatiotemporal gene expression, TFs determine cell fate during development and cellular responses to the environment. TFs are classified into structural families based on previously described DNA-binding domains (DBDs) (Weirauch and Hughes, 2011). In eukaryotes, over 80 families of TFs are recognized, and the list includes homeodomains (e.g. Hox genes), nuclear receptors (ESR1; estrogen receptor) and P53 (TP53; tumor suppressor). Additional families of TFs are believed to exist (Weirauch and Hughes, 2011). The structure of a TF contains at least one DBD. However, they can have more than one DBD and additional regulatory domains (Frietze and Farnham, 2011). Therefore, TFs and their regulation have a global effect on cellular phenotypes, highlighting the need for their identification. Advances in DNA sequencing technologies have greatly increased the number of available genome sequences, enabling researchers to interrogate gene expression, TFs determine cell fate during development and cellular responses to the environment. TFs are classified into structural families accord-
3 Benchmark

We benchmarked CREPE’s performance to manually curated human TF datasets. We applied the TF Cataloguing function of CREPE to the human proteome (GRCh38.p13, Ensembl 107) (Cunningham et al., 2022). The proteome was pre-processed to only include the primary isoform per gene [from 120 712 to 23 486 sequences], and sequences derived from alternative mappings were removed [from 23 486 to 20 409 sequences]. We identified 1519 genes (~7.4% of the input sequences) as putative TFs across 51 families (Supplementary Table S1). As expected, the five largest TF families were the C2H2 zinc fingers (zf-C2H2), homeodomains, basic helix–loop–helix (bHLH), basic leucine zippers (bZIP) and forkhead (Fig. 1).

To validate our findings, we compared them to curated human TF datasets: CisBP 2.00 (Weirauch et al., 2014), which was pre-processed to include TFs with known TF DBD [from 1639 to 1546 genes] and the Human Transcription Factors catalog (hTFcat) (Lambert et al., 2018), which was also pre-processed to remove TFs with unknown DBDs [from 1639 to 1570 genes]. CREPE retrieved 91.6% (1416/1546) of the TFs in CisBP and 90.2% (1416/1570) of the TFs in the hTFcat (Fig. 2A and B). Comparison by family shows that 28 of the 51 (55%) TF families identified by CREPE showed parity (Fig. 2C; Supplementary Table S2).

CREPE identified 103 genes as TFs across 21 families, which were not included in the reference TF datasets (Supplementary Fig. S1A and Supplementary Table S2). The largest family included in this group (21/103; 20%) are the high-mobility group box, and includes HMGB1-4, TOX and TOX2-4 which to date, have not yielded DNA-binding specificity motifs (Lambert et al., 2018). The second largest family is the C2H2 zinc fingers with 20 genes, of which 11 are annotated by Ensembl as novel and 3 as read-throughs (Cunningham et al., 2022) (Supplementary Fig. S1B and Supplementary Table S2). To understand the generalities of the CREPE-only group, we performed Gene Ontology (The Gene Ontology Consortium, 2021) analysis and found overrepresentation of epigenetic terms, such as histone H3-K4 demethylation (GO:0034720), histone H3-K9 trimethylation (GO:0036124) and histone H4-K20 methylation (GO:0034770) (Supplementary Table S3), and contains genes such KDM5A, ARID4B, NSD2 whose debate of whether they are sequence specific TFs is ongoing (Lambert et al., 2018). However, this group also contains genes such as HSFX3, HSFX4, DUXB, FOXL3, KLF18 that are members of known TF families. Whether they possess any DNA-binding sequence-specificity and TF activity needs to be validated experimentally.

CREPE missed 130 genes that are in both reference human TF datasets, spanning 16 families (Supplementary Fig. S2). The largest family in this group accounting for over half of the genes (79/130; 60.8%) are C2H2 zinc fingers. A notable difference between CREPE and the references appeared in the AT Hook and C2H2 zinc finger families. This was expected because it is known that current models can make the classification of C2H2 zinc fingers and AT...
Hook DBDs difficult (Lambert et al., 2018). Taken together, our strategy using CREPE to identify the human TFs is summarized with a true positive rate of 90.2%.

4 Application

We applied CREPE to catalog the TF repertoire of Heliconius melpomene and Heliconius erato, two species of butterflies that are found throughout Central and South America and are a popular model system to study phenotypic determination through their wing coloration patterning mimicry (McMillan et al., 2020). We obtained predicted proteomes for both species from LepBase (Challi et al., 2016). Next, we applied the CREPE TF Cataloguing function and catalogued 664 and 599 TFs in H.melpomene and H.erato, respectively (Supplementary Fig. S5). In both species, TFs were distributed across 51 families. The top three families in both butterflies are zf-C2H2, homeodomain and bHLH, accounting for over half of the identified putative TFs. In preparation to run the CREPE TF Annotation function, we generated the required putative TF gene trees using OrthoFinder (Emms and Kelly, 2019) and 20 animal proteomes from Ensemble (Cunningham et al., 2022) (Supplementary Table S4). These gene trees were then used as inputs for the TF Annotation function. The annotations of the putative TFs for H.melpomene and H.erato mapped to Drosophila melanogaster are available in our GitHub.

5 Conclusion

In this work, we described CREPE, a Shiny app to systematically catalogue and annotate TFs. CREPE was benchmarked against curated human TFs datasets, obtaining a 90.2% true positive rate. Using CREPE, we identified putative TFs that are currently not listed in curated references, suggesting that by using CREPE a user can obtain an updated TF catalog. The intention behind this tool is to allow researchers to explore the TFs in poorly annotated genomes or from understudied organisms. To showcase this functionality, we catalogued the TF repertoire of H.melpomene and H.erato butterflies. Taken together, CREPE provides a path forward for large-scale TF identification. A tutorial on how to execute CREPE and all data underlying this article can be found in our GitHub (github.com/dirostri/CREPE) and in its online supplementary material.

Acknowledgements

We would like to thank Dr. Steven Van Belleghem for his insights and support in this project and Dr. Humberto Ortiz-Zuazaga for computational support.

Author contributions

Diego A. Rosado-Tristani (Conceptualization [lead], Data curation [equal], Formal analysis [lead], Methodology [equal], Software [lead], Writing—original draft [lead], Writing—review & editing [equal]) and Jose Arcadio Rodriguez-Martinez (Conceptualization [equal], Data curation [supporting], Formal analysis [supporting], Funding acquisition [lead], Investigation [equal], Methodology [supporting], Project administration [lead], Writing—original draft [supporting], Writing—review & editing [equal]).

Funding

D.A.R.-T. was supported by the RISE Fellowship 3R2GM061151-21. This work was supported by NSF [1736026], NIH [SC1GM127231], University of Puerto Rico Rio Piedras Institutional Funds (FPI) and the NIH Institutional Development Award (IDeA) INBRE [P20GM103475].

Conflict of Interest

none declared.

References

Challi, R. et al. (2016) Lepbase: the Lepidopteran genome database. bioRxiv, 56994.
Cunningham, F. et al. (2022) Ensembl 2022. Nucleic Acids Res., 50, D998–D999.
Eddy, S. (2011) Accelerated profile HMM searches. PLoS Comput. Biol., 7, e1002195.
Emms, D.M. and Kelly, S. (2019) OrthoFinder: phylogenetic orthology inference for comparative genomics. Genome Biol., 20, 238.
Friesz, S. and Farnham, P. (2011) Transcription factor effector domains. In: Hughes, T.R. (ed.) A Handbook of Transcription Factors. Springer, Dordrecht, Heidelberg, London, New York, pp. 261–278.
Hotaling, S. et al. (2021) Toward a genome sequence for every animal: where are we now? Proc. Natl. Acad. Sci. USA, 118, e2109019118.
Lambert, S. et al. (2018) The human transcription factors. Cell, 172, 650–665.
McMillan, W. et al. (2020) From patterning genes to process: unraveling the gene regulatory networks that pattern heliconius wings. Front. Ecol. Evol., 8, 1–15.
Mistry, J. et al. (2021) Pfam: the protein families database in 2021. Nucleic Acids Res., 49, D412–D419.
Paradis, E. and Schliew,K. (2019) ape 5.0: an environment for modern phylogenetics and evolutionary analyses in R. Bioinformatics, 35, 526–528.
The Gene Ontology Consortium. (2021) The gene ontology resource: enriching a GOld mine. Nucleic Acids Res., 49.
Weirauch, M. and Hughes, T.R. (2011) A catalogue of eukaryotic transcription factor types, their evolutionary origin, and species distribution. In: Hughes, T.R. (ed.) A Handbook of Transcription Factors. Springer, Dordrecht, Heidelberg, London, New York, pp. 25–74.
Weirauch, M. et al. (2014) Determination and inference of eukaryotic transcription factor sequence specificity. Cell, 158, 1431–1443.
Wittkopp, P.J. and Kalay, G. (2011) Co-regulatory elements: molecular mechanisms and evolutionary processes underlying divergence. Nat. Rev. Genet., 13, 59–69.