

dms2dfe: Comprehensive Workflow for Analysis of Deep Mutational Scanning Data

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Software

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Summary

`dms2dfe` is an integrative analysis workflow designed for end-to-end analysis of Deep Mutational Scanning (Fowler, Stephany, and Fields 2014) data. Using this workflow, users can implement various processing methods and downstream applications for a pairwise enrichment analysis of ultra-deep sequencing data.

Recently, owing to the evolution of sequencing and phenotyping technologies, large scale genotype to phenotype data is increasingly being generated. Along this line of research, Deep Mutational Scanning method allows comprehensive assessment of all possible amino acid substitutions of an entire gene or part of a gene. In the analysis of Deep Mutational Scanning data, `dms2dfe` addresses crucial issue of noise control using widely used DESeq2 (Love, Huber, and Anders 2014) workflow and offers variety of downstream analyses to contextualize results. In downstream analyses, `dms2dfe` workflow provides identification of potential molecular constraints, comparative analysis across different experimental conditions and generation of data-rich visualizations (Dandage and Chakraborty 2016). While a number of tools have been developed for analysis of DMS data (Fowler et al. 2011; Bloom 2015; Rubin et al. 2017), users familiar with commonly used state-of-art genomics tools such as Trimmomatic (Bolger, Lohse, and Usadel 2014), Bowtie (Langmead and Salzberg 2012), samtools (Li et al. 2009) and DESeq2 (Love, Huber, and Anders 2014) can opt for `dms2dfe` workflow for analysis of preferential enrichments. Note that `dms2dfe` workflow is designed exclusively for experimental designs in which there is a need of pair-wise analysis of samples eg. before and after selection.

As an input for the workflow, deep sequencing data (whether unaligned or aligned) or list of genotypic variants can be provided. For a demonstration purpose, sample datasets from various studies (Firnberg et al. 2014; C. A. Olson, Wu, and Sun 2014; Melnikov et al. 2014) are available here.¹ `dms2dfe` uses DataFrames from robust Pandas library (McKinney 2010) for processing all the tabular data. For enabling downstream analyses, from user-provided PDB file, structural features are extracted (Kabsch and Sander 1983; Sanner, Olson, and Spehner 1996) and conservation scores are obtained from multiple sequence alignments (Sievers and Higgins 2014; Pupko et al. 2002). As an optional step, visualizations of preferential enrichments onto PDB structure are generated using UCSF Chimera (Pettersen et al. 2004).

Source code and issue tracker is available in `dms2dfe`'s GitHub repository.² Documentation and API³ are generated using Sphinx.⁴

¹https://github.com/rraadd88/ms_datasets

²<https://github.com/kc-lab/dms2dfe>

³<https://kc-lab.github.io/dms2dfe>

⁴<http://www.sphinx-doc.org>

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