MMTF 2017 Hackathon

Summary Slides June 28, 2017

Finding Structurally Similar Proteins (Kevin Savage)

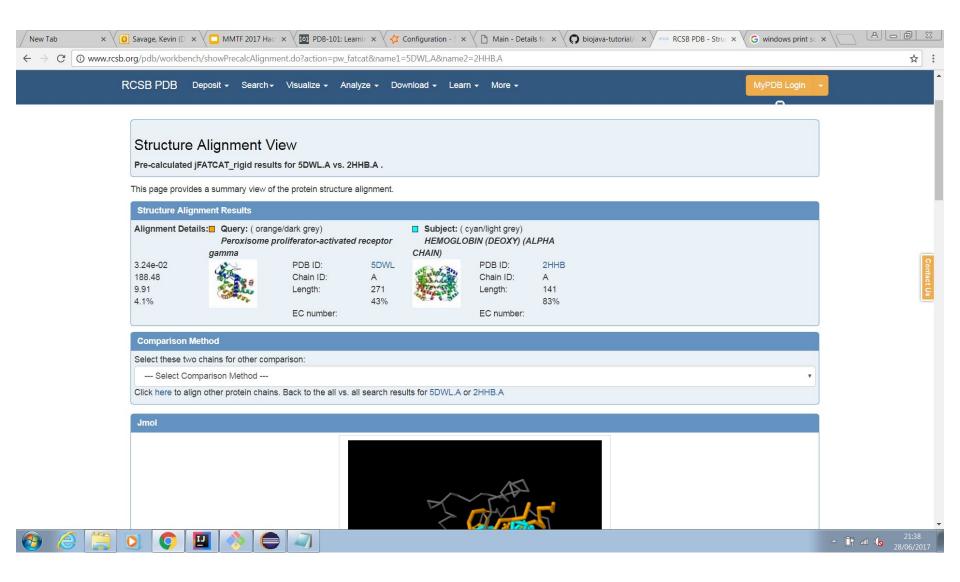
- Goal: Find something in the PDB that is structurally similar to a new protein (e.g. that has just come off the beamline) with 2HHB as example.
- Used Biojava to calculate similarity, Spark to run aver all PDB:
 - AFPChain chain = algorithm.align(atoms1, atoms2);
 - AFPChainScorer.getTMScore(chain, atoms1, atoms2)
- E.g. output:

4JYH.C was similar 0.9804168929359692

1IRD.A was similar 0.9805141907275625

5DWL.B was similar 0.9985411517813645

5DWL vs 2HHB (one of 122 from 1/10th of the PDB)



PDBj Mine 2 filter feature (I)

- MMTF doesn't contain the full metadata
- Sometimes you want to filter using stuff that isn't in MMTF, but is in mmCIF
- PDBj's Mine 2 RDB holds all mmClF data and more (except atom data ⇒ use MMTF for that!), all accessible via a powerful SQL interface
- Issues:
 - The service is very flexible, so the column names are not predefined (cannot be hardcoded)
 - By default the `pdbid` column is used for MMTF filtering, can be overwritten by the user
 - If you want to filter by pdbid.chainid, you can define it on SQL level and then inform the API to use that column for filtering
 - Newlines aren't handled properly by Spark's CSV parser (fixed in 2.2)

PDBj Mine 2 filter feature (II)

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```
public static void main( String[] args ) throws IOException
    // goal: use an sql query to get a list of pdbids, then filter the MMTF-DB to only include those entries
    SparkConf conf = new SparkConf().setMaster("local[*]").setAppName(App.class.getSimpleName());
    JavaSparkContext sc = new JavaSparkContext(conf);
    String path = System.getProperty("MMTF_FULL");
        System.err.println("Environment variable for Hadoop sequence file has not been set");
        System.exit(-1);
    // read PDB in MMTF format
    JavaPairRDD<String, StructureDataInterface> pdb = MmtfReader.readSequenceFile(path, sc);
    String sql = "select pdbid from brief summary where modification date >= '2016-06-28' and resolution < 1.5";
    MineSearch search = new MineSearch(sql);
    System.out.println("Number of entries in MMTF 10% library matching query #1: "+pdb.filter(search).keys().count()+"/"+search.dataset.count());
    // Retrieve PDB chain sequences matching to the Pfam accession "PF00046" (Homeobox)
    // and having a resolution better than 2.0 Angstrom and a sequence length greater than or equal to 58 (residues)
    // https://pdbj.org/mine/sql?query=SELECT+concat(s.pdbid%2C+%27.%27%2C+s.chain)+as+structureChainId%2C+s.*%2C+r.ls d res high+as+reso%2C%0A+++++LENGTH(p.pdbx seq one letter code can)+as+len%2
    sql = "SELECT concat(s.pdbid, '.', s.chain) as \"structureChainId\", s.*, r.ls_d_res_high as reso,"+
                "LENGTH(p.pdbx_seq_one_letter_code_can) as len, " +
                "('>' || s.pdbid || s.chain) as header, " +
               "replace(p.pdbx_seq_one_letter_code_can,E'\n','') as aaseq " + // the replace here is required because spark's csv parser cannot handle newlines properly (should be fixed in 2.2)
            "FROM sifts.pdb_chain_pfam s " +
            "JOIN refine r on r.pdbid = s.pdbid " +
            "JOIN entity_poly p on p.pdbid = s.pdbid " +
            "AND s.chain = ANY(regexp_split_to_array(p.pdbx_strand_id,',')) " +
            "WHERE pfam_id = 'PF00046' " +
                "AND r.ls_d_res_high < 2.0 " +
            "AND LENGTH(p.pdbx_seq_one_letter_code_can) >= 58 " +
            "ORDER BY reso, len, s. chain ";
    MineSearch search2 = new MineSearch(sql, "structureChainId", true);
    search2.dataset.show(10);
    System.out.println("Number of entries in MMTF 10% library matching query #2: "+pdb.flatMapToPair(new StructureToPolymerChains()).filter(search2).keys().count()+"/"+search2.dataset.count());
    sc.close();
```

PDBj Mine 2 filter feature (III)

```
Number of entries in MMTF 10% library matching query #1: 188/1857
structureChainId|pdbid|chain|sp_primary|pfam_id|reso|len|header| aaseq|
                                   P56178 | PF00046 | 1.85 | 65 | > 4rduA | GHMVRKPRTIYSSFQLA... |
           4rdu.A| 4rdu|
                                   P56178 | PF00046 | 1.85 | 65 | > 4rduD | GHMVRKPRTIYSSFQLA... |
           4rdu.D| 4rdu|
                                    P02836 | PF00046 | 1.9 | 61 | > 2hddA | MAEKRPRTAFSSEQLAR... |
           2hdd.A 2hdd
                                    P02836 | PF00046 | 1.9 | 61 | > 2hddB | MAEKRPRTAFSSEQLAR... |
           2hdd.Bl 2hddl
           2hos.A 2hos
                                    P02836 | PF00046 | 1.9 | 63 | > 2hosA | GSDEKRPRTAFSSEQLA... |
                                   P02836 | PF00046 | 1.9 | 63 | > 2hosB | GSDEKRPRTAFSSEQLA... |
           2hos.Bl 2hosl
                             В
                                    P40424 | PF00046 | 1.9 | 73 | >1 pufB | ARRKRRNFNKQATEILN... |
           1puf.B | 1puf |
           1puf.A 1puf
                                   P09631 | PF00046 | 1.9 | 77 | > 1 pufa | NNPAANWLHARSTRKKR... |
                                   P14859 | PF00046 | 1.9 | 161 | >1e3oC | EEPSDLEELEQFAKTFK... |
           1e3o.C| 1e3o|
           3cmy.A 3cmy
                                   P23760 | PF00046 | 1.95 | 61 | >3cmyA | GORRSRTTFTAEQLEEL... |
                                   -----
only showing top 10 rows
```

Number of entries in MMTF 10% library matching query #2: 2/12

Importing validation reports (Przemek Porebski)

Not all residues in crystal structures are created equal. Data analysis should take into account that some residues may not be supported by electron density (erroneously modeled) or be disordered more than others.

The level of "experimental support" can be defined as a correlation between the density calculated from the model and experimental density. Currently this data is calculated by PDB during deposition of the structure and was calculated for all structures that have deposited structure factors. The values as part of validation reports which are available as XML files.

E.g.

http://files.rcsb.org/pub/pdb/validation_reports/gm/2gmo/2gmo_validation.xml.gz

The goal is to get this data available for filtering residues with poor fit to electron density using MMTF-Spark

Importing validation reports (Przemek Porebski)

Approach 1: Use XML files directly as data source

- a. Use additional library: https://github.com/databricks/spark-xml
- b. Problem: Structure of validation reports is not convenient for using as a data source
- c. Resulting schema:

```
-- NatomsEDS: long (nullable = true)
|-- _altcode: string (nullable = true)
|-- _avgoccu: double (nullable = true)
|-- _chain: string (nullable = true)
|-- _cis_peptide: string (nullable = true)
|-- _corrupt_record: string (nullable = true)
|-- _ent: long (nullable = true)
|-- _icode: string (nullable = true)
|-- _model: long (nullable = true)
|-- _num-H-reduce: long (nullable = true)
|-- _owab: double (nullable = true)
|-- phi: double (nullable = true)
|-- psi: double (nullable = true)
|-- _rama: string (nullable = true)
|-- _resname: string (nullable = true)
|-- _resnum: long (nullable = true)
|-- _rota: string (nullable = true)
-- _rscc: double (nullable = true)
|-- _rsr: double (nullable = true)
|-- _rsrz: double (nullable = true)
|-- _said: string (nullable = true)
|-- _seq: long (nullable = true)
|-- clash: array (nullable = true)
      |-- element: struct (containsNull = true)
           |-- _VALUE: string (nullable = true)
|-- _atom: string (nullable = true)
|-- _cid: long (nullable = true)
           |-- _clashmag: double (nullable = true)
            -- _dist: double (nullable = true)
```

Importing validation reports (Przemek Porebski)

Approach 2: Convert XML files to more convenient, curated data structure and dump as parquet files.

```
|-- altcode: string (nullable = true)
|-- chain: string (nullable = true)
|-- clash: double (nullable = true)
|-- icode: string (nullable = true)
|-- model: long (nullable = true)
|-- resname: string (nullable = true)
|-- resnum: long (nullable = true)
|-- rscc: double (nullable = true)
|-- rsr: double (nullable = true)
|-- structureId: string (nullable = true)
|-- ligRSRZ: double (nullable = true)
```

Conversion script:

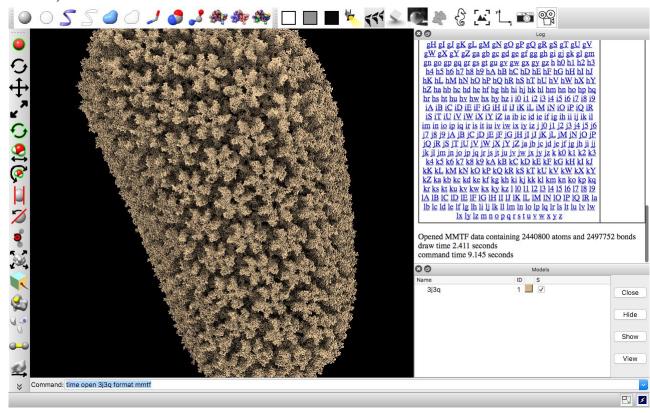
https://gist.github.com/hokinus/5c033fe8f8fcff539b595ae071da337d

For analysis filter out polymer residues with rsrz < 2 (poor fit to electron density or high disorder).

Small molecules with ligRSRZ > 5. (called LLDF) are suspicious, because their fit to electron density is significantly worse than surrounding residue. May indicate lack of evidence for a ligand

Port ChimeraX MMTF Reader to C++

- Greg Couch, UCSF RBVI, http://www.rbvi.ucsf.edu/chimerax/
- Opens 3j3q in ~8 seconds when using C++, didn't wait to find out how slow
 Python version was (mmCIF is ~14 seconds)
- Overhead in Python version was due to creating Python objects for every atom, bond, etc.



MMTF Feature Request (for Chimera)

- Separate data structure for
 - metal center bonds from mmCIF
 - split out metal bond in chemical components
 - H-bonds from mmCIF
- MMTF spec: add info about case of strings, e.g., entityType
- Add example code to traversal examples that shows how to map chain ids to entities (needed to find out which chains are polymers so a representation of gaps in the structure can be made)

• **Goal**: find small protein fragments with similar conformation to a query fragment structure.

Solution:

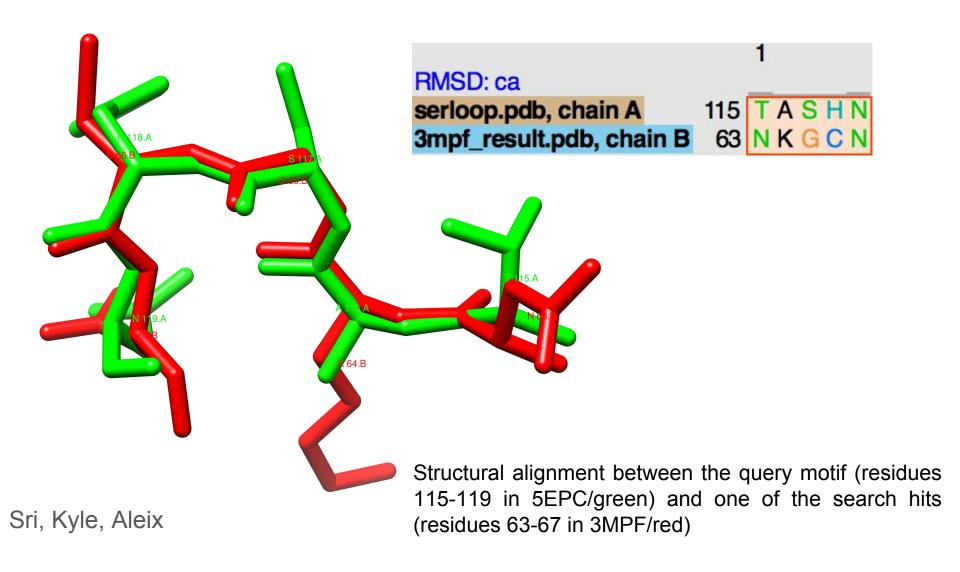
- a. Create a flatMap function to extract all possible fragments of a polypeptide chain.
- b. Create a similarity measure to compare the fragments with the query.

Problems:

a. BioJava objects are not serializable.

```
// Quick hack, the user has to take care of providing that
Group[] query = (Group[]) StructureIO.getStructure("~/Downloads/5epc_fragment.pdb")
        .getChainByIndex(0).getAtomGroups().toArray(new Group[5]);
Double[] phi = new Double[query.length - 1];
Double[] psi = new Double[query.length - 1];
for (int i = 0; i < query.length - 1; i++) {
    double phi_q = Calc.getPhi((AminoAcid) query[i], (AminoAcid) query[i + 1]);
    double psi_q = Calc.getPsi((AminoAcid) query[i], (AminoAcid) query[i + 1]);
    phi[i] = phi_q;
    psi[i] = psi_a;
```

| ++ | +- | + | | | | | | |
|----------------------------|----|----------------------------|--|--|--|--|--|--|
| l pdblchainlresnuml scorel | | | | | | | | |
| + | +- | + | | | | | | |
| 15EPC1 | ΑI | 115 6.267985741978399E-7 | | | | | | |
| 15EPC1 | ВΙ | 115 0.026708508240375108 | | | | | | |
| 13HM21 | Εl | 154 0.02710099780993716 | | | | | | |
| 14T011 | Αl | 46210.0278050219621015331 | | | | | | |
| IBIBWI | ΑI | 9910.0292424648755662331 | | | | | | |
| I5HCD1 | ΑI | 1653 0.029461464080928607 | | | | | | |
| 15HCC1 | ΑI | 1653 0.030920538529235358 | | | | | | |
| I3MPFI | ВΙ | 63 0.031284202410822234 | | | | | | |
| I4BQVI | ВΙ | 18600.031292878387673216 | | | | | | |
| 1ESE | ΑI | 255 0.031396152324964274 | | | | | | |



NGL viewer for matching multiprotein complexes (Markus Wiederstein)

GOAL: Speed up visualization of superposition results from TopMatch/TopSearch for multiprotein complexes

(http://topsearch.services.came.sbg.ac.at)

Get two structures ("query" and "target") into NGL stage.

Status: done

2. Get structure alignment from TopMatch and apply transformation matrix to target.

Status: done

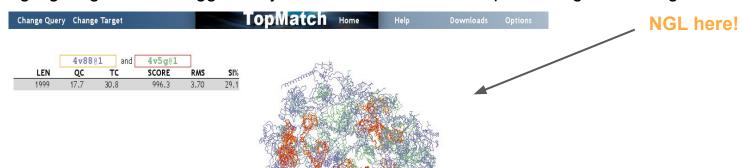
3. Get list of equivalent residues from TopMatch and color respective parts.

Status: done for single chain comparisons, but

problem: access to label_asym_id for selecting chains in multiprotein complexes

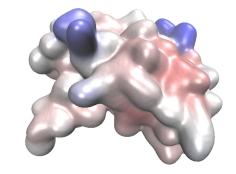
Next steps:

- Switch from NGL.ColormakerRegistry.addSelectionScheme
- MMTF writer (Python/C++) for uploaded structures
- Highlighting residues triggered by mouse-over events in sequence alignment widget

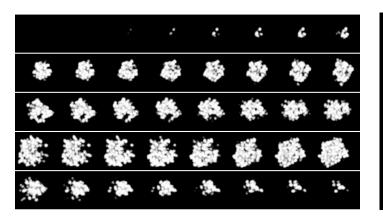


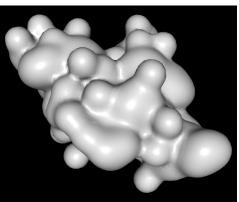
Fast Molecular Surfaces for NGL Viewer

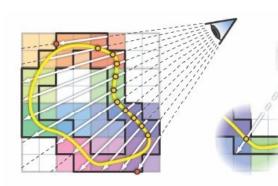
(Alexander Rose, Michael Krone)

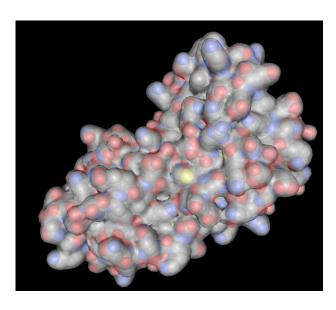


- Volumetric molecular surfaces for NGL Viewer (similar to QuickSurf in VMD)
 - Use WebGL / GPU for fast computation
 - Use REGL API (https://github.com/regl-project/regl)
- Try different rendering methods
 - Marching Cubes isosurfaces / Surface Nets
 - Direct Volume Rendering (isosurface ray marching)
- Missing features / Work in Progress
 - Coloring + volume-based Ambient Occlusion
 - Volume Ray Marching
 - Improve computation & rendering speed
 - Extend this to Solvent Excluded Surfaces?









Biopython Structure → MMTF file

(Nathan, Yiling)

Problem

 Biopython can read MMTFs but not write them

Solution

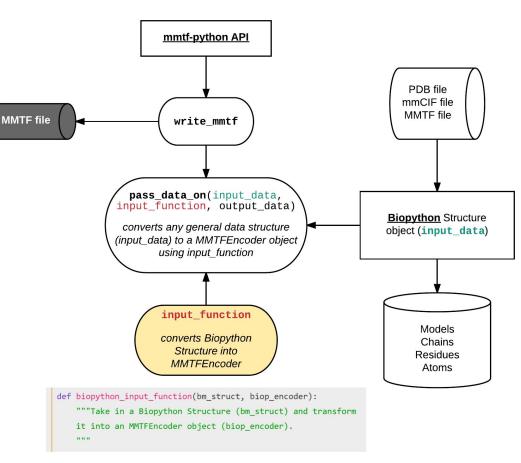
- Existing mmtf-python API can accept custom write solutions for any data type
- Write custom function to convert
 Biopython Structure to MMTFEncoder
 object

Problems

- mmtf-python API incomplete, untested
- Able to write but not load completed MMTF file

Missing features

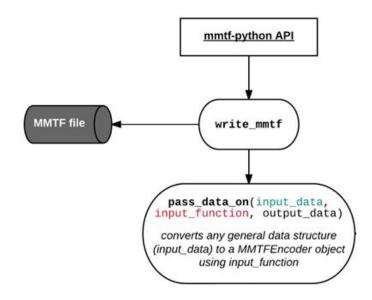
Biopython does not parse
 bioassembly, header, bond counts, etc
 to do!



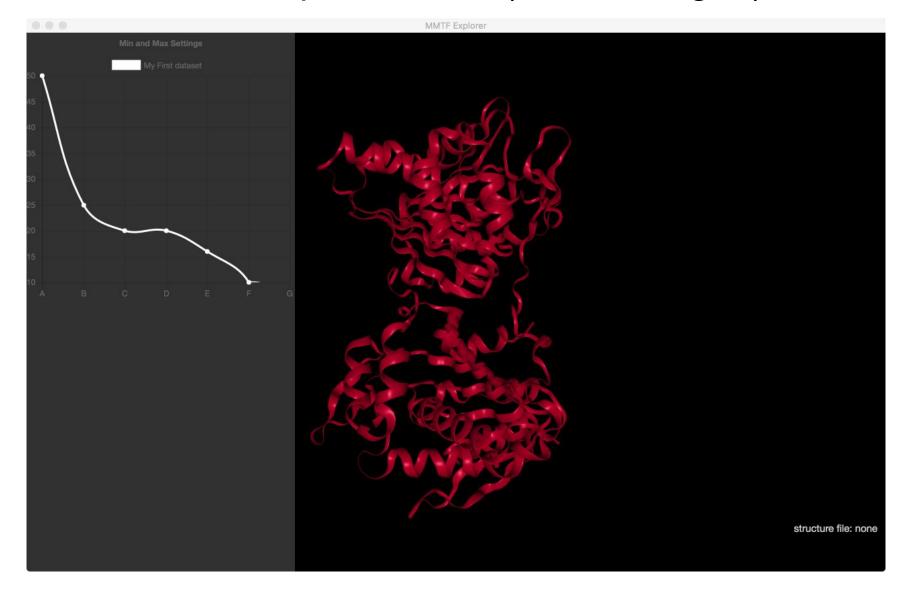
Write MMTF file

(Nathan, Yiling)

- Problem
 - MMTF-python does not write files
- Solution
 - Existing mmtf-python API can accept custom write solutions for any data type
 - Write custom function to write MMTF files
- Problems
 - mmtf-python API incomplete, untested
- Missing features
 - Atoms and groups



SAXS Client to Spark Server (S. Doutreligne)



A 'diff' utility for macromolecular coordinates Wolfram Tempel

- problem: even insignificant changes in coordinates would confound text-based comparison
- general use case: highlight significant differences within pairs of structures
- specific use case: automate annotation of model revisions (Thanks to P. Porebski)
- simplest case: aligned models only differ in coordinates
- added complexity: unaligned models, inconsistent residue ids, types, etc.
- natural potential for parallelization: for each atom in structure 1, find corresponding atom in structure 2, based on distance and atom type/context after alignment
- desired output: list of inter-model differences, sorted by "significance": missing/additional atoms, mutated residue types, translated coordinates

Identify structural Variation based on SNPs

Alexander seitz

Mapping of IDs:

ftp://ftp.uniprot.org/pub/databases/uniprot/current_release/knowledgebase/idmapping_ing/idmapping_selected.tab.gz

Genes (Output Format: gtf)

http://genome.ucsc.edu/cgi-bin/hgTables

(switch "Track" to "TransMapEnsembl")

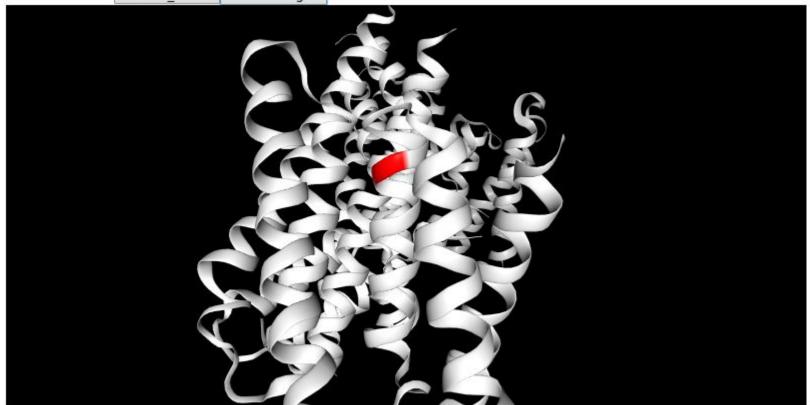
```
chr20 43534683 . A G . . chr20 43534803 . T C . .
```

Choose Files testInput.vcf_modified.vcf
available IDs: 2C23:A_180 ▼ colorChanges



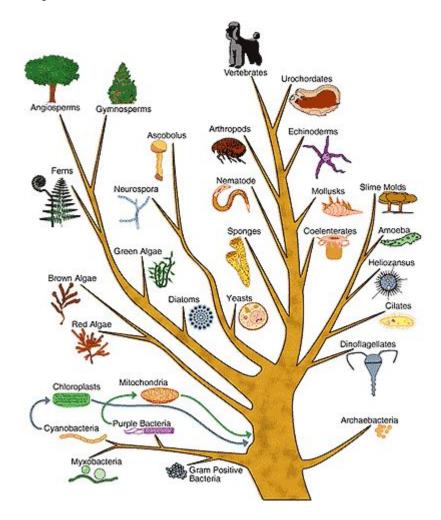
Choose Files testInput.vcf_modified.vcf

available IDs: 2C23:A_180 ▼ colorChanges



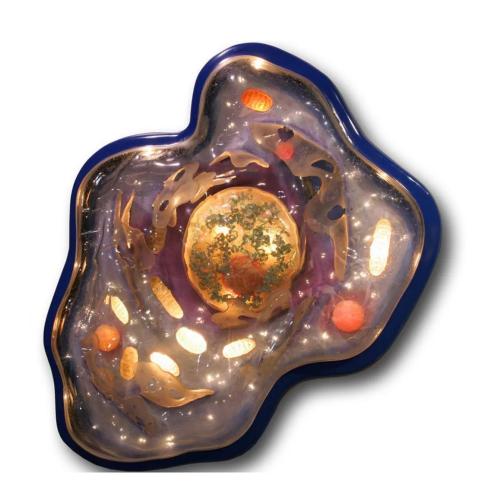
Evolution Explorer

- Goal: Analytics friendly Data
 Frame organization of PDB data
 with initial emphasis on evolution
- First, organize structures by domains of life
- Next, organize by enzyme classification, transporter classification, GO ontologies, etc.
- Filter structures to only those that have a minimum number of samples from each domain



Evolution Perspective

- Scientists interested in early evolution need samples from archaea, bacteria, and eukarya.
- It would be nice to have at least two of each (certainly at least one)
- How many proteins and nucleic acid structures have that kind of diversity?
- What kinds of properties do these samples have in common?
- Is there a statistically significant difference for some properties?



Code Snippet

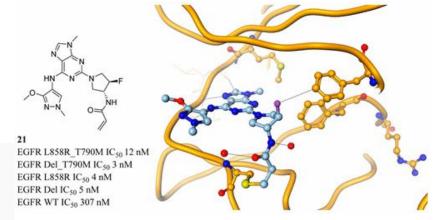
- Major hurdle was just set up. Learning Java, Spark, and everything else.
- Successfully have three datasets in "Data Frame" format, that can be unioned, reduced, aggregated, etc.
- Next step would be to group structures by identity
- Then, collect additional data of interest

1. E.g. EGFR L858R vs EGFR Mutant

http://www.rcsb.org/pdb/explore/explore.do?structureId=5U8L http://www.rcsb.org/pdb/explore/explore.do?structureId=5UG8

- 2. https://github.com/jjgao/mutantpdb
- 3. https://jjgao.github.io/mutantpdb/

| EGFR | P00533 | S | 645 C | S645C | Likely Oncogenic Likely Gain—of—function |
|------|---------|---|--------|-------|---|
| EGFR | P00533 | Н | 773 L | H773L | Oncogenic Gain-of-function |
| EGFR | P00533 | G | 810 S | G810S | Likely Oncogenic Gain-of-function |
| EGFR | P00533 | L | 858 Q | L858Q | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | N | 826 S | N826S | Oncogenic Gain-of-function |
| EGFR | P00533 | L | 858 M | L858M | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | Н | 773 Y | H773Y | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | Р | 596 L | P596L | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | K | 467 T | K467T | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | G | 724 S | G724S | Oncogenic Gain-of-function |
| EGFR | P00533 | R | 836 C | R836C | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | V | 774 A | V774A | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | L | 861 Q | L861Q | Oncogenic Gain-of-function |
| EGFR | P00533 | R | 776 H | R776H | Likely Oncogenic Gain-of-function |
| EGFR | P00533 | G | 719 S | G719S | Oncogenic Gain-of-function |
| EGFR | P00533 | G | 735 S | G735S | Oncogenic Gain-of-function |
| EGFR | P00533 | N | 826 Y | N826Y | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | L | 703 P | L703P | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | L | 747 V | L747V | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | G | 598 V | G598V | Oncogenic Gain-of-function |
| EGFR | P00533 | Ĺ | 861 P | L861P | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | Ē | 734 0 | E7340 | Likely Oncogenic Gain-of-function |
| EGFR | P00533 | Ē | 868 G | E868G | Likely Oncogenic Gain-of-function |
| EGFR | P00533 | Ē | 709 G | E709G | Likely Oncogenic Likely Gain-of-function |
| EGFR | P00533 | Ē | 746 V | E746V | Likely Oncogenic Likely Gain-of-function |
| LUIN | 1 00333 | _ | 7 TO V | L/40V | Likety oncogenite Likety dain-of-function |



Matt Hudson: https://github.com/mhudson-compbio/

Alexey Strokach: https://github.com/ostrokach/

Daniel Kool: https://github.com/kool7d/

Yana Valasatava: https://github.com/valasatava
Jamaine Davis: https://github.com/jdavislab/

JJ Gao: https://github.com/jjgao

- Goal: Compare structure of single AA mutants
- Steps involved
 - Get a list of oncogenic mutations
 - Map to PDB positions
 - Extract residues from PDB
 - NGL view of mutant to WT structures
- Problems encountered
 - Acquiring the amino acids in the structures based on pdb id, chain and residue number
 - Would be great if MMTF interface can support
 - Error handling when failed to load a structure in Spark

Matt Hudson: https://github.com/mhudson-compbio/

Alexey Strokach: https://github.com/ostrokach/

Daniel Kool: https://github.com/kool7d/

Yana Valasatava: https://github.com/valasatava
Jamaine Davis: https://github.com/idavislab/

JJ Gao: https://github.com/jjgao

- Use information from mutations.parquet to select required structures
- Create pdb id and pdb chain id HashSets to filter the structures and chains in MMTF FULL
 - downloadMmtfFiles missing many structures (e.g. 4ZWH, 4ZWK, 5BPR, 5BNU)
 - HashSet works much faster than list (as expected)
- Get 3621 MMTF chains in ~1 second

| iprot Ref Position Varaint chainId insCode pdbAtomPos pdbId | JavaPairRDD <string, structureda<br="">.readSequenceFile(path, .filter(t -> pdb_ids.co .flatMapToPair(new Stru .filter(t -> pdb_chains</string,> |
|---|--|
| 04637 E 326 L A null 26 1A1U | |
| 04637 F 328 V A null 28 1A1U | |
| 04637 T 329 I A null 29 1A1U | |
| 04637 L 330 R A null 30 1A1U | |
| 04637 L 330 H A null 30 1A1U | Matt Hudson: |

```
ataInterface> chains = MmtfReader
. sc)
ontains(t. 1))
uctureToPolymerChains())
s ids.contains(t. 1));
```

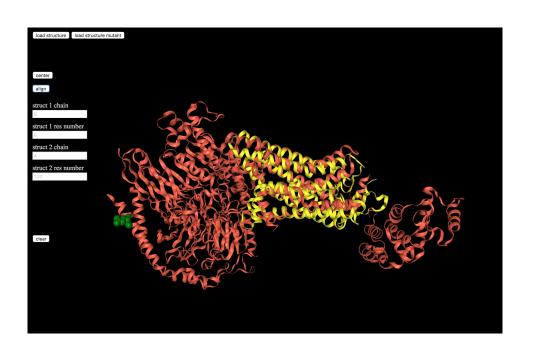
ttps://github.com/mhudson-compbio/

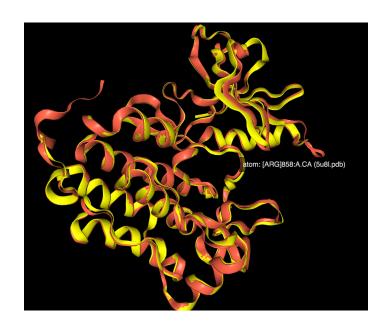
n: https://github.com/ostrokach/ os://github.com/kool7d/

a: https://github.com/valasatava https://github.com/jdavislab/

github.com/jigao

https://jjgao.github.io/mutantpdb/





Matt Hudson: https://github.com/mhudson-compbio/

Alexey Strokach: https://github.com/ostrokach/

Daniel Kool: https://github.com/kool7d/

Yana Valasatava: https://github.com/valasatava
Jamaine Davis: https://github.com/jdavislab/

JJ Gao: https://github.com/jjgao

Hydrogen Bonds Dataset (Luiz Borro)

Create a dataset with all hydrogen bonds for a given list of PDB structures

- Three types of Hydrogen Bonds
 - Main Chain Main Chain
 - Main Chain Side Chain
 - Side Chain Side Chain

| str | uctureId ch | + ain1 re | + sidue1 at | om1 ele | + ment1 in | dex1 ch | + ain2 re | +- sidue2 a | + tom2 ele | ment2 in | dex2 distance Int | eractionType |
|-----|-------------|--------------|----------------|---------|---------------|---------|--------------|----------------|---------------|----------|-----------------------|----------------------|
| + | 10GS | + | + TYR | + | + | + 11 | + | PROI | 0l | + | 0 2.2522461 | + |
| ŀ | 10GS | A A | THR | N I | N I | 2 | A A | TYR | 0 | 0 | 1 2.254425 | HBond_MM HBond_MM |
| ì | 10G5 | Al | THR | NI | N | 2 | Al | GLN | 0 | Ol | 54 3.0278215 | HBond MM |
| î | 10GS | A | VAL | N | N | 3 | A | THR | O | oj | 2 2.2466044 | HBond MM |
| Î | 10GS | A | VAL | N | N | 3 | A | LYS | 0 | O | 27 2.927684 | HBond_MM |
| | 10GS | A | VAL | N | N | 4 | A | VAL | 0 | 0 | 3 2.2475886 | HBond_MM |
| | 10GS | A | ARG | N | N | 11 | A | TYR | OH | 0 | 5 3.0740633 | HBond_MS |

Amino Acids Contacts (Luiz Borro)

Creates a dataset with all hydrogen bonds for a set of protein structures

- Internal and Interface Contacts
- Hydrogen Bonds
 - 1. Main Chain Main Chain
 - 2. Main Chain Side Chain
 - 3. Side Chain Side Chain
 - Aromatic Contacts
 - Charged Interactions
 - Hydrophobic Interactions
 - Disulfide bonds

Water-mediated hydrogen bond?

Amino Acids Contacts (Luiz Borro)

| t | uctureId ch | + ain1 re | siduellat | om1lele | ment1 in | dex1 ch | ain2 re | sidue2la | tom2 ele | ment2lir | ndex2 distance Int | eractionTyne |
|---|-------------|--------------|-----------|---------|----------|---------|---------|----------|----------|----------|---------------------|--------------|
| + | | + | + | + | + | + | + | +- | + | + | + | + |
| 1 | 10GS | A | TYR | N | N | 1 | A | PRO | 0 | 0 | 0 2.2522461 | HBond_MM |
| Ì | 10GS | A | THR | N | N | 2 | A | TYR | 0 | 0 | 1 2.254425 | HBond MM |
| Î | 10GS | A | THR | N | N | 2 | A | GLN | 0 | 0 | 54 3.0278215 | HBond MM |
| Î | 10GS | A | VAL | N | N | 3 | A | THR | 0 | 0 | 2 2.2466044 | HBond MM |
| Î | 10GS | A | VAL | N | N | 3 | A | LYS | 0 | 0 | 27 2.927684 | HBond MM |
| Ì | 10GS | A | VAL | N | N | 4 | A | VAL | 0 | 0 | 3 2.2475886 | HBond MM |
| | 10GS | A | ARG | N | N | 11 | A | TYR | OH | 0 | 5 3.0740633 | HBond_MS |