# Data organization: modeling and normalization

Brad Chapman Ginkgo Bioworks

28 March 2019

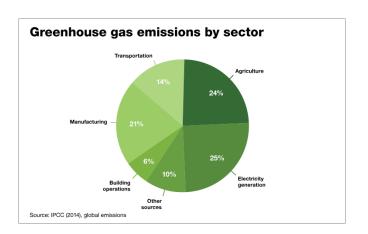


https://www.ginkgobioworks.com/

#### Motivation



### Agriculture



#### https:

//www.gatesnotes.com/Energy/We-should-discuss-soil-as-much-as-coal



# Argicultural fertilizer usage



### **JoynBio**

#### ON JOYN BIO

We design microbes for more sustainable agriculture.

Technology ......

Every tempoon of soil is home to billions of microbes

By engineering the bacteria that live in soil, we aim to nurture healthy and sustainable communities—from the ground up.

https://joynbio.com/

#### Animal proteins

#### **ENVIRONMENTAL IMPACT**

- 18% of global anthropogenic greenhouse gas emissions come from livestock farming. By contrast, global transportation accounts for 13%.<sup>1</sup>
- 26% Earth's ice-free surface is used for livestock farming.
   This represents 70% of all agricultural land.<sup>2</sup>
- 27-29% of humanity's freshwater footprint is used for the production of animal products.<sup>3</sup>
- Livestock farming is a top contributor to deforestation, land degradation, water pollution and desertification.4

https://www.new-harvest.org/about

### Motif Ingredients

#### EVERYONE SEES A DIFFERENT FUTURE FOR FOOD

#### WE'RE MAKING INGREDIENTS FOR **EVERYONE**.



Motif makes ingredients for the next generation of plant-based and healthy foods. We use fermentation to brew vital proteins and nutrients that power your body and please your palate. We collaborate with chefs, health experts, and food visionaries to create new building blocks for tomorrow's food revolution. Motif combines tradition, innovation, and biotechnology to help build a more sustainable future, for everyone.

### Synthetic Biology as a solution



https://grist.org/series/panic-free-gmos/

# What does Ginkgo do?

- Make organisms that produce products
- Multiple hosts: bacteria, yeasts, mammalian cells
- Many projects: find activity, improve enzymes, make proteins.

# Organization of making

- Organism Engineers: define a project goal
- Design: what to make
- Build: the designed constructs
- Test: find best performing constructs

#### Design: what to make

- Source input enzymes and pathways
  - Existing biological knowledge
  - Permutations of known enzymes
- Adjust to be possible to make
  - Codon optimization
  - Restriction site/design issues
- Transcriptional machinery

#### Build: how to make it

Order parts: https://www.twistbioscience.com/

- Synthesize difficult parts: internal BioFab (<- Gen9 <- Codon)
- Combine parts into pathways: internal assembly approaches

#### Test: how well did we make it

- Develop high throughput assays for compound of interest
- Scaling
  - Samples (plate-based)
  - Size (fermentation)
- Analysis: combining results from multiple inputs

### Data challenges

- Multiple levels of highly specialized work
- Custom designs, builds, assays required
- Lots of simultaneous projects with different needs

#### Data goals

- Capture intent and process in a lightweight way
- Data interoperability
- Enable permisionless data analysis
- Inform decision making

#### Data reality





#### Roles and Responsibilities:

- Automate horrible business practices
- Write ad hoc SQL as needed

#### REQUIRED EXPERIENCE:

- 15 years exp deep learning in Python
- PhD thesis on Bayesian modeling
- NLP experience in 7 languages
- 10 years of creating Hadoop clusters from scratch

9:18 PM - 11 Feb 2019

https://twitter.com/NickDoesData/status/1095160141207531520



# Do science, not cleaning





Just a personal anecdote, but, in the past 2 years, % of any given project:

- + that involves ML: 15%
- + that involves moving, monitoring, and counting data to feed ML: 85%

8:34 AM - 15 Jan 2019

https://twitter.com/vboykis/status/1085183529527508992

#### Interoperable structured data is hard



### Analogous to workflows

```
#!/usr/bin/env cwl-runner
cwlVersion: v1.0
class: Workflow
doc: BwA-mapping-PE is a mapping workflow using BwA for Peared-end reads.

inputs:
    fastq1_url:
        type: string
    label: Download link of FastQ file from next generation sequencer
    fastq2_url:
        type: string
    label: Download link of FastQ file from next generation sequencer

module load bwa/9.7.15

bwa mem -t 16 -M /clusterfs/rosalind/users/makman/GATK/bwa mem/HanXROr1.0-20151230.fa
```

Anzac\_Pueblo\_R1\_trimmed.fastq.gz Anzac\_Pueblo\_R2\_trimmed.fastq.gz > Anzac\_Pueblo.sam

/HanXROr1.0-20151230.fa Arikara R1 trimmed.fastg.gz Arikara R2 trimmed.fastg.gz >

# bwa mem -t 16 -M /clusterfs/rosalind/users/makman/GATK/bwa mem

Arikara.sam

### Common issue everywhere

- Most knowledge in documents, presentations, scientists's brains
- A lot of work to represent specialized knowledge in a structured way
- We need context in new biological areas

### Data approach

Without imposing too much extra work:

- Capture scientific intent
- Improve naming with ontologies

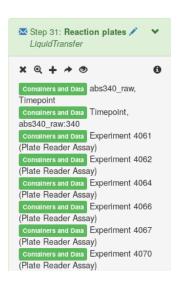
### Scientific intent: challenges

- High level data structures to organize projects
- Multiple ways of doing design and test
- Ad-hoc capture of intent: Jupyter, Excel, Slack

### Scientific intent: approach

- Examine characteristic analyses
- Model uncaptured data
- Provide data structures

#### Example: lab workflow



### Example: analysis setup

```
# Container IDs
uninduced_plate = range(97852, 97852+4)+[97849]
uninduced_od_plate = range(98235, 98235+5)
induced_plate = range(98286, 98286+5)
induced_od_plate = range(98401, 98401+5)
reaction_plate = range(99665, 99665+5)
glycerol_plate = range(95062, 95062+5)
```

#### Example: analysis output

#### 11 Files Out

```
In [459]: # Writes dataframes into multi-sheet excel files

def save_xls(list_dfs, sheet_names, xls_path):
    writer = pd.ExcelWriter(xls_path)
    for i, df in enumerate(list_dfs):
        df.to_excel(writer, sheet_names[i])
    writer.save()

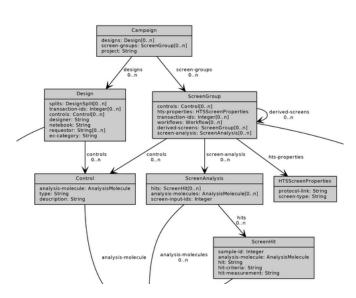
write_directory = '/ginkgo/bitome/data/users/ashepard/PythonData/'

In [460]: list_dfs = [df, posdf, negdf, condf, cont_stat, df10, df11, df12, df13, dfall]
    sheet_names = ['All_Data', 'Positive Control', 'Negative Control', 'WT Control',
```

### Example: missing connections

- Experimental intent
- Design to assay connection
- Assay to analysis connection
- Analysis output hits to reporting

#### Data model view



#### Flexible data model

```
^{:graphviz/color "gray80"}
ScreenGroup
[^{:type Control :cardinality [0 n]} controls
 ^HTSScreenProperties hts-properties
 ^{:type Integer :cardinality [0 n]} transaction-ids
 ^{:type Workflow :cardinality [0 n]} workflows
 ^{:type ScreenGroup :cardinality [0 n]} derived-screens
 ^{:type ScreenAnalysis :cardinality [0 n]} screen-analysis
^{:graphviz/color "gray80"}
ScreenHit
[^Integer sample-id
^AnalysisMolecule analysis-molecule
^String hit
^String hit-criteria
^String hit-measurement
```

https://github.com/luchiniatwork/hodur-engine

# Translating into database storage



#### Flexible

Datomic's provides the power of schema without requiring that you define everything up-front. Add attributes dynamically at any time, without worrying about fixed tables or disruptive migrations. Learn More »



#### Hierarchical

A universal relation lets you handle row-oriented, column-oriented, graph, and hierarchical data in a single system without impedance. <u>Learn</u> More »

https://www.datomic.com/



### Ontologies: extend key/values

- Need
  - Standard naming
  - Flexibility
- Adopt existing ontologies
  - Avoid work of inventing
  - Contribute and extend community standards

#### Adoption: descriptions and assays

- Feature descriptions: Sequence Ontology
  - http://www.sequenceontology.org/
- HTS: BioAssay Ontology
  - http://bioassayontology.org/
- Mapping to Gene Ontology
  - http://geneontology.org/docs/ download-mappings/

#### Example key names: OD600

```
600

OD600

OD600:600

abs600_raw
```

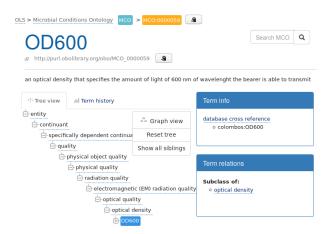
### BioAssay Ontology



https://www.ebi.ac.uk/ols/ontologies/bao



### Microbial Conditions Ontology



#### Other issues

- Where to put units
- Which mean the same thing?

```
Time
Timepoint
Timepoint (second)
Timestamp
```

# Initial steps: data mine from existing

- Need to make practical
- Extract examples
- SciGraph
- Semi-automatically map to ontologies

```
https://github.com/ginkgobioworks/ontology-clean
```

### SciGraph



https://github.com/SciGraph/SciGraph

#### Search example

```
$ curl 'http://localhost:9000/scigraph/vocabulary/search/time' \
  | json_pp
 "definitions" : [
 "A unit which is a standard measure of the dimension "
 "in which events occur in sequence."],
"labels" : ["time unit"],
"iri": "http://purl.obolibrary.org/obo/U0_0000003"
},
 "definitions" : [
  "A quality in which events occur in sequence."],
"labels" : ["time"],
"iri": "http://purl.obolibrary.org/obo/PATO_0000165"
```

### RDF like modeling

Ontologies map naturally to flexible entity - attribute - value

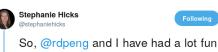
```
assay1 - sample - sample1
assay1 - sample - sample2
sample1 - time - 30
sample1 - time unit - seconds
sample1 - OD600 - 0.986
sample2 - time - 30
sample2 - time unit - seconds
sample2 - OD600 - 1.13
```

# Table like modeling

Many tools work with tabular formats, but have to transform arbitrary columns into final analysis tables

assay	sample	time	time unit	OD600
assay1	sample1	30	seconds	0.986
assay1	sample2	30	seconds	1.13

#### Elements and Principles of Data Science



So, @rdpeng and I have had a lot fun recently thinking about data analyses and the field of #datascience as a whole. We wrote down some ideas in a #preprint called Elements and Principles of Data Analysis. feedback is welcomed #statistics #AcademicTwitter arxiv.org /abs/1903.07639

3:18 PM - 20 Mar 2019

https://arxiv.org/abs/1903.07639 https://twitter.com/stephaniehicks/status/1108462768099856384



### Framework for organizing analyses

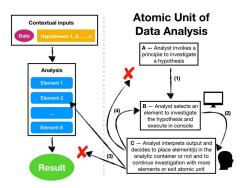


Figure 8: The atomic unit of data analysis. Given the contextual inputs, the data analysis is built upon atomic units of data analyses, which occurs both on the computer and in the analyst's head. In each atomic unit, the data analyst first chooses a principle to investigate a hypothesis or scientific question (Stage 4a). Then, the analyst alternates between Stages B and C until the analyst exits the atomic unit, either by choosing to end the line of investigation or choosing to invoke a new principle. The actions in the atomic unit are denoted in by (1) the analyst selects an element to investigate a hypothesis or scientific question, which is executed in the computer or console, (2) the analyst interprets the output in his or her head, (3) the analyst decides whether or not to place the element from (1) in the data container or data product or not, which means the element is never recorded in the data container or product and the audience does not see it as part of the analysis, and (4) the analyst decides to whether to continue in the atomic unit by selecting another element to investigate the hypothesis or question or to exit the atomic unit entirely and end this line of investigation.

# In conclusion: machine learning

```
from sklearn import linear model
first300 = joined[joined['bucket'] < 300]
x300 = first300['bucket'].values.reshape(-1,1)
v300 = first300['ratio'].values.reshape(-1.1)
regr = linear_model.LinearRegression()
regr.fit(x300, y300)
first300.plot(x='bucket',y='ratio',kind='scatter')
plot.plot(first300['bucket'], regr.predict(x300), color='blue', linewidth=3)
[<matplotlib.lines.Line2D at 0x7f6371a23cf8>]
  0.18
  0.16
  0.14
  0.10
  0.08
  0.06
  0.04
  0.02
  0.00
```

#### Real conclusion

- Ginkgo: synthetic biology for agricultural sustainability
- General data challenges
- Capture scientific intent
- Improve naming with ontologies