# A data pipeline approach to orphan disease insights

- KNIME 2018 FALL SUMMIT -

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#### **Key Themes in Artificial Intelligence in Rare Disease Diagnosis**

2018

Rady Children's Institute

**Impact** 

A new world record of 19.5 hours is set for the fastest genetic diagnosis by Rady Children's Institute for Genomic Medicine.

http://www.frontlinegenomics.com/review/21973/65-years-of-dna/

**Computational Learning** 



 Computational hypothesis generation, data interpretation, decision support, and acceleration of human insight to enable rare disease diagnosis

Manage bias

**Information Fusion** 



 Semantically integrate and navigate complex, heterogeneous, local and distributed data Manage ambiguity

"Big Data"



Capture, process, filter, and manage a global and growing avalanche of internal and external scientific and clinical data

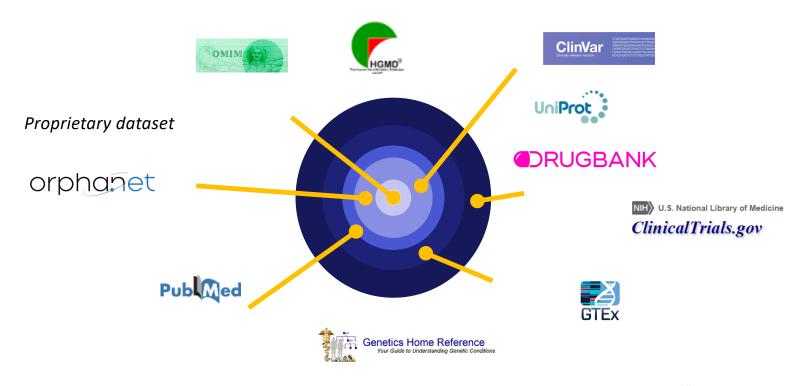
Manage variety





## The Alexion Insight (AI) Engine

#### MAPPING AND INTEROGATING THE UNIVERSE OF RARE DISEASES



Each link represents a set of evidences that are extracted using a complex data pipeline







# The Alexion Insight (AI) Engine

#### MAPPING AND INTEROGATING THE UNIVERSE OF RARE DISEASES

#### All Rare Diseases

► There are approximately 9,500 rare diseases reported in Orphanet, OMIM, and GHR — all captured our data graph

#### Incidence/Prevalence

- Mine registry and genetic information
- ► Insight into disease tractability
- ► Patient segmentation

#### Disease Onset

 Clarify window of opportunity and identify patient populations

#### Genetic Architecture

▶ Detailed mapping of all known genetic mutations in the target disease with proprietary bioinformatics

#### **5** Competitive Intensity

► Estimate competitive landscape via our novel scoring scheme using drugs, clinical trials, medical specialties...

#### Severity and unmet needs

- ► Find the most severe diseases via phenotype ontology and age of death
- ► Standard of care

Each link represents a set of evidences that are extracted using a complex data pipeline







## The Alexion Insight (AI) Engine

#### ANSWERING KEY RARE DISEASE QUESTIONS





orphanet



UniProt 3



Clinical Trials.gov



**Proprietary datasets** 

XML, JSON, CSV, RestAPI, SQL...

\*Edward Tufte on visual design of a data analysis

Enforce appropriate comparisons\*



Inspect & evaluate alternative explanations\*

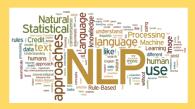
Recognize the

nature of the

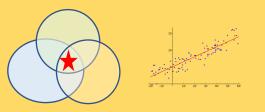
inherent multivariate

problem\*

Document the source and characteristic of the data\*



Demonstrate mechanism of cause & effect and express them quantitatively\*



Each question is answered with a dedicated data analytics pipeline highlighting key insights



Conditions of interest

Clinical asset deal flow

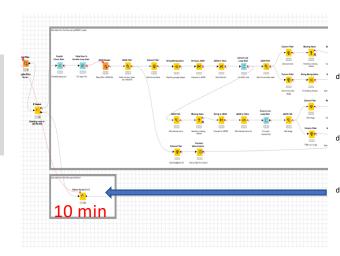
Orphan disease diagnostics

**KOL** mapping

Patient population size and segmentation

# JSON parsing example

of complex JSON files



Next: Big Data extension → Spark JSON?

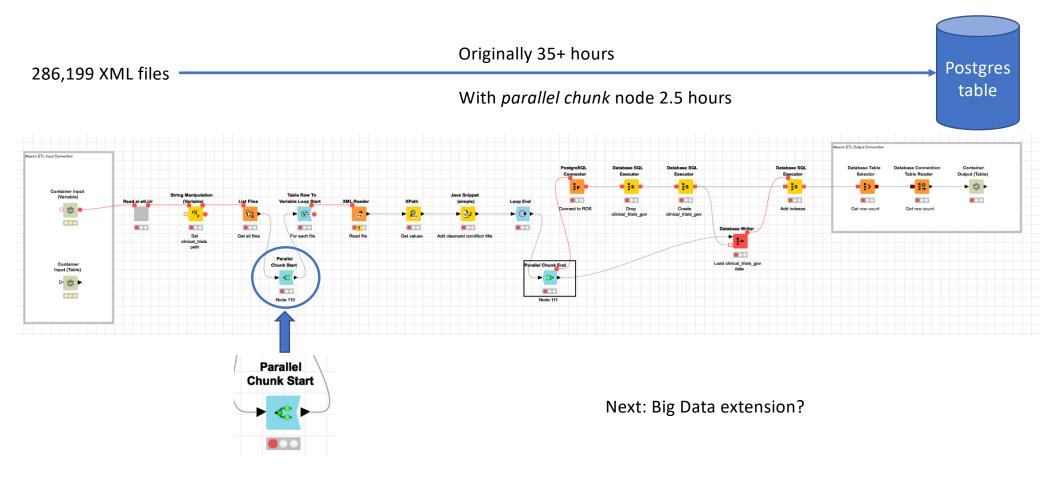
```
import pandas
import json
import glob
def get_clinical_trials(filename):
    with open(filename) as drugbank_file:
       drugbank_data = json.load(drugbank_file)
       drugbank_id = drugbank_data.get('drugbank_id')
       trials = drugbank_data.get('clinical_trials')
       clinical_trials = []
          for trial in trials:
              sponsors = trial.get('sponsors')
              if not sponsors:
                  add_clinical_trial(clinical_trials, drugbank_id, trial)
                   for sponsor in sponsors:
                       add_clinical_trial(clinical_trials, drugbank_id, trial, sponsor)
def get all clinical trials():
   for \ filename \ in \ glob.iglob (flow\_variables['drugbankplus\_path'] + '/**/DB*.json', \ recursive=True):
       clinical_trials.extend(get_clinical_trials(filename))
    return clinical trials
def get_drug_value(value_name, interventions):
   return set([drug_name for intervention_drug_names in [[drug.get(value_name) for drug in intervention.get('drugs')]
                                                       for intervention in interventions]
                for drug_name in intervention_drug_names])
def add_clinical_trial(clinical_trials, drugbank_id, trial, sponsor = None):
   conditions = trial.get('conditions')
   condition_titles = '|'.join(set([condition.get('title') for condition in conditions])) if conditions else None
   condition synonyms = '|'.join(filter(None, [('|'.join(condition.get('synonyms')) if condition.get('synonyms') else None)
                                             for condition in conditions])) if conditions else None
   interventions = trial.get('interventions')
   intervention kinds = '|'.join(set([intervention.get('kind') for intervention in interventions])) if interventions else None
   intervention_titles = '|'.join(set([intervention.get('title') for intervention in interventions])) if interventions else None
   drug names = '|'.join(get drug value('name', interventions)) if interventions else None
   drug\_drugbank\_ids = '|'.join(get\_drug\_value('drugbank\_id', interventions)) \ if interventions \ else \ None
   clinical\_trials.append(\{'drugbank\_id': drugbank\_id, 'identifier': trial.get('identifier'), 'status': trial.get('status'), identifier'), 'status': trial.get('status'), identifier': trial.get('identifier'), 'status': trial.get('status'), identifier'), 'status': trial.get('status'), identifier': trial.get(
                          'title': trial.get('title'), 'official_title': trial.get('official_title'), 'purpose': trial.get('purpose'),
                          'phases': '|'.join(map(str, trial.get('phase'))) if trial.get('phase') else None,
                          'start date': trial.get('start date'), 'end date': trial.get('end date'),
                           'brief_summary': trial.get('brief_summary'),
                           'condition_titles': condition_titles, 'condition_synonyms': condition_synonyms,
                          'intervention kinds': intervention kinds, 'intervention titles': intervention titles,
                           'drug names': drug names, 'drug drugbank ids': drug drugbank ids,
                          'sponsor title': sponsor.get('title') if sponsor else None,
                           'sponsor_agency_class': sponsor.get('agency_class') if sponsor else None,
                          'lead_sponsor': sponsor.get('lead_sponsor') if sponsor else None})
```

**Postgres** 

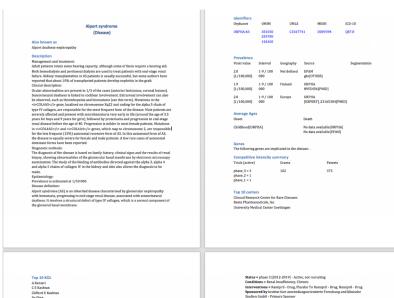
table

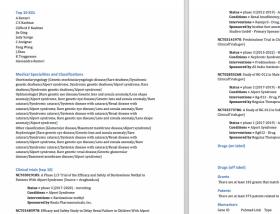
output\_table = pandas.DataFrame(get\_all\_clinical\_trials())

# Clinical trials example

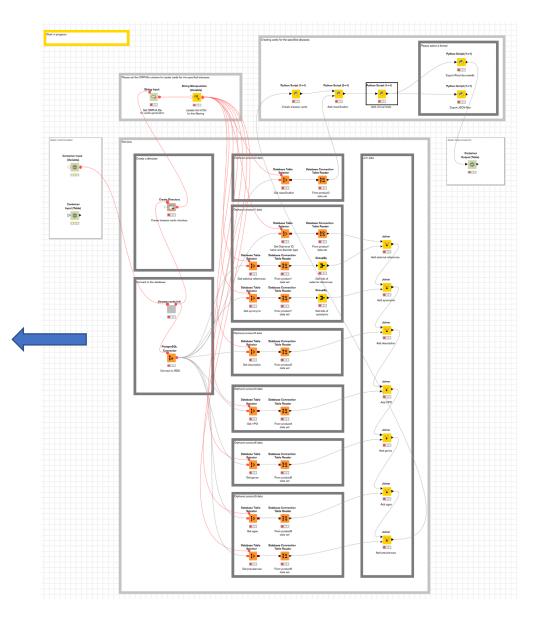


# Disease card example











## Insights must be FAIR\*

#### Findable, Accessible, Interoperable, Reusable

Package your insight with a version:

- **Description of the question/problem** with assumptions/inclusions (e.g. presentation, publication)
- A readable data analysis story supporting the insight (e.g. KNIME workflow, *jupyter* notebook)
- Data/Model snapshot used for the analysis (e.g. python pickle file, flat file export, Model file)
- Any supporting dataset (e.g. gene-disease mapping, knowledgebase timestamp, training set)
- Code used to run the analysis (e.g. label in GitHub/Artifactory, python virtualenv)
- Environment that ran the code (e.g. AWS AMI, python virtualenv)
- Use/re-use templates/packages (e.g. KNIME metanode, Artifactory, python virtualenv)

You will want to know how you came up with a given insight Your Commercial, BD or Corp Strategy partner will ask!...trust me

Place the "Description of the analysis" inside one of your enterprise document repository so your work can be found by enterprise tools like *Search* and *ExpertFinder*.

\* https://www.go-fair.org/fair-principles/

