

Automated Hypothesis Generation Based on Mining Scientific Literature

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San Jose, California

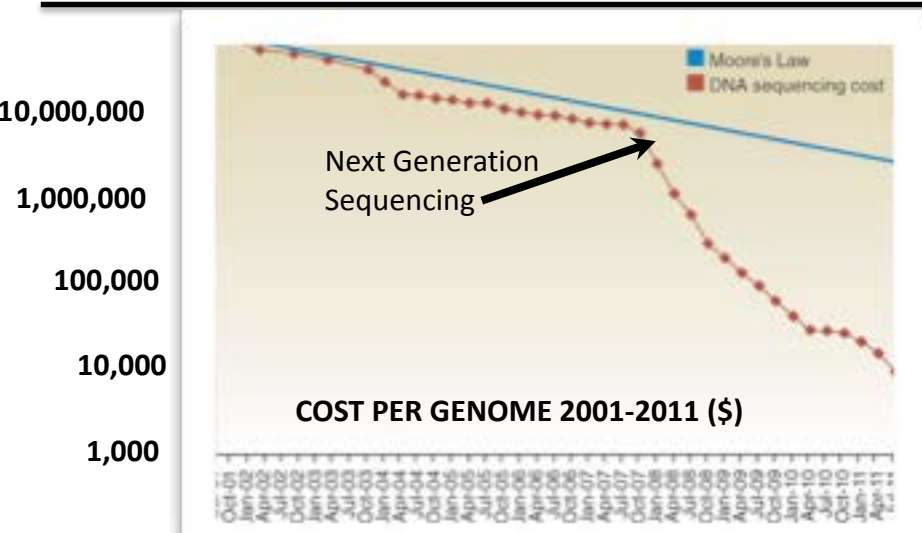
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³Baylor College of Medicine

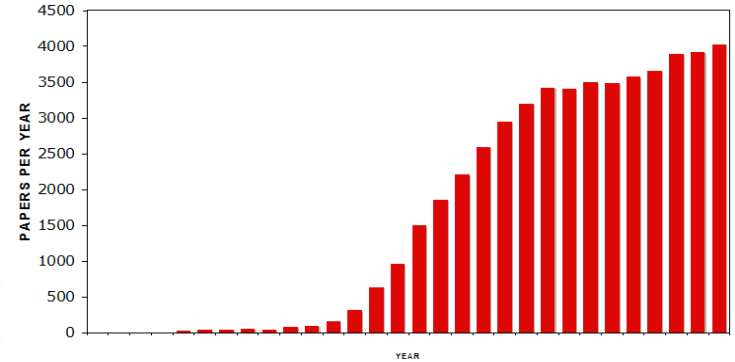
Houston, Texas

McNair Medical Institute, DARPA

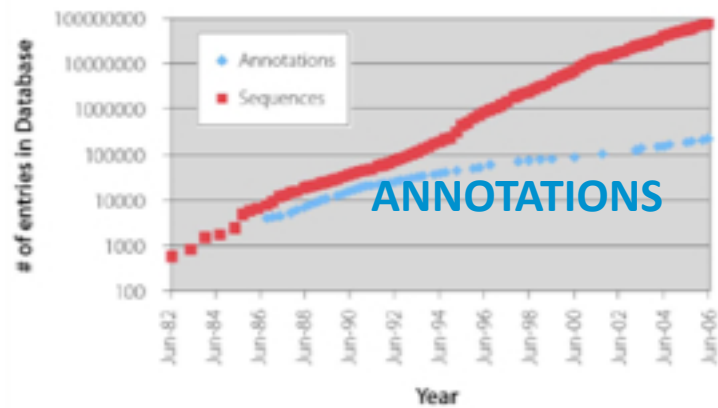
DATA OVERFLOW



Yearly distribution of new p53 papers

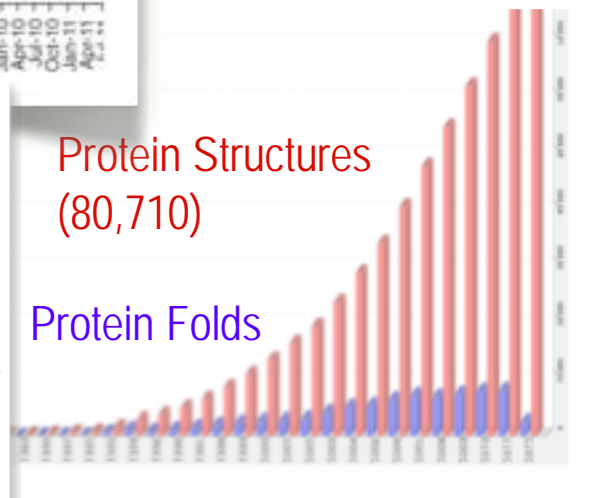


Growth of sequences and annotations since 1982



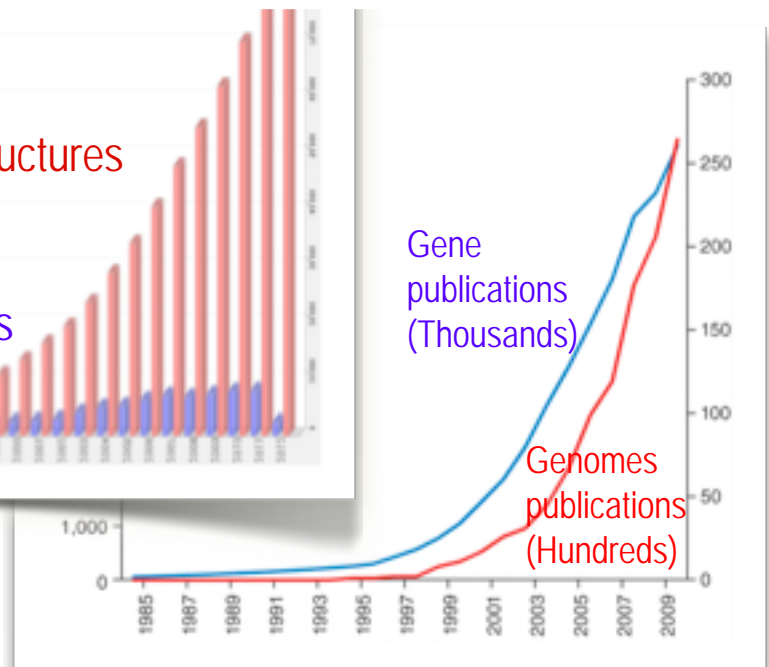
Protein Structures
(80,710)

Protein Folds



Gene
publications
(Thousands)

Genomes
publications
(Hundreds)



A mismatch between raw data and our analytic abilities

PUBLICATION OVERFLOW

Overall

- 50 million scientific papers
- 1 million more per year
- 2 new papers per minute

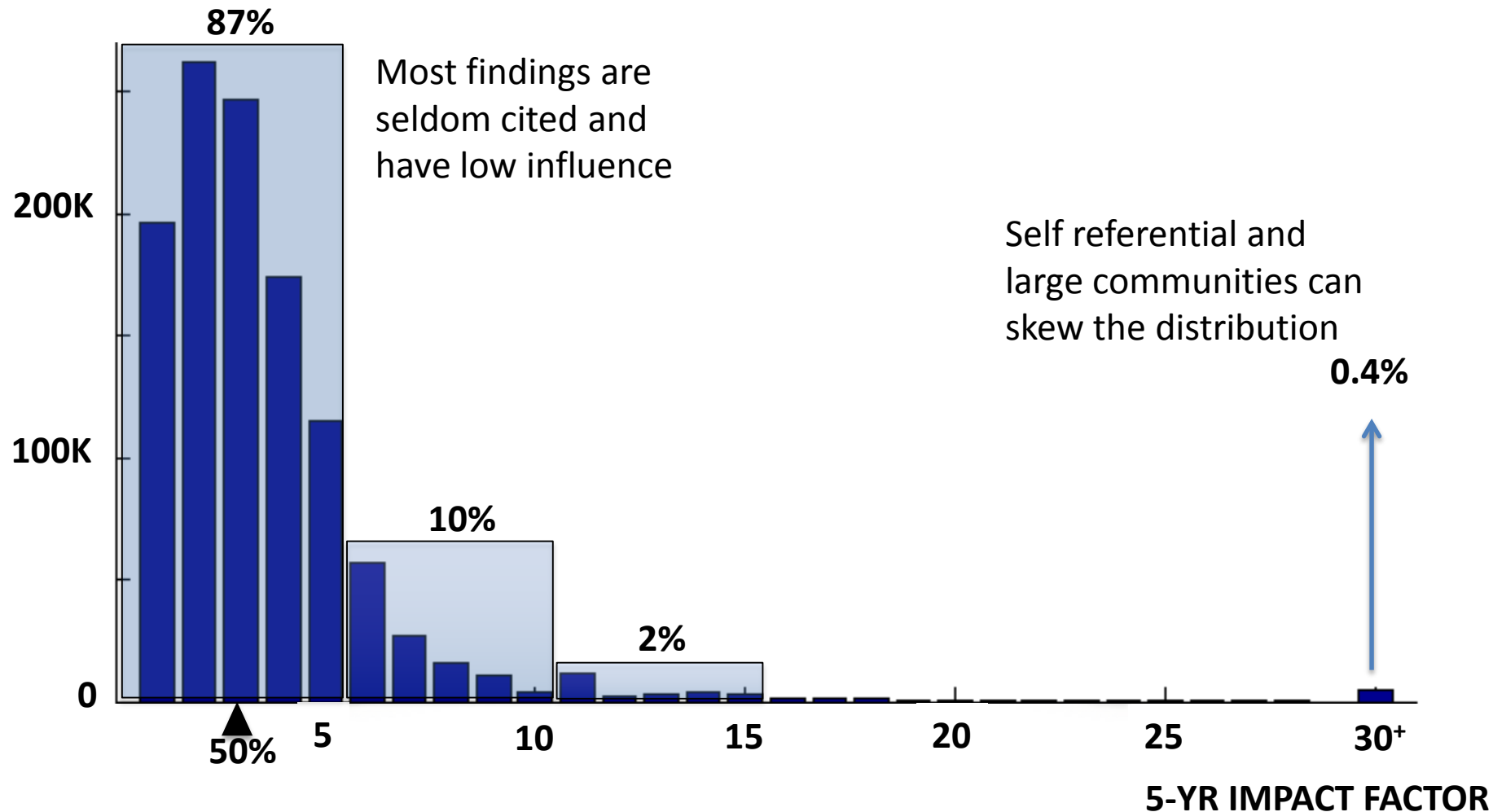
Biomedical research

- 10^3 to 10^5 papers per topic areas
- Over 70,000 papers on p53 (a tumor suppressor)

A fundamental bottleneck: we cannot keep up with discovery

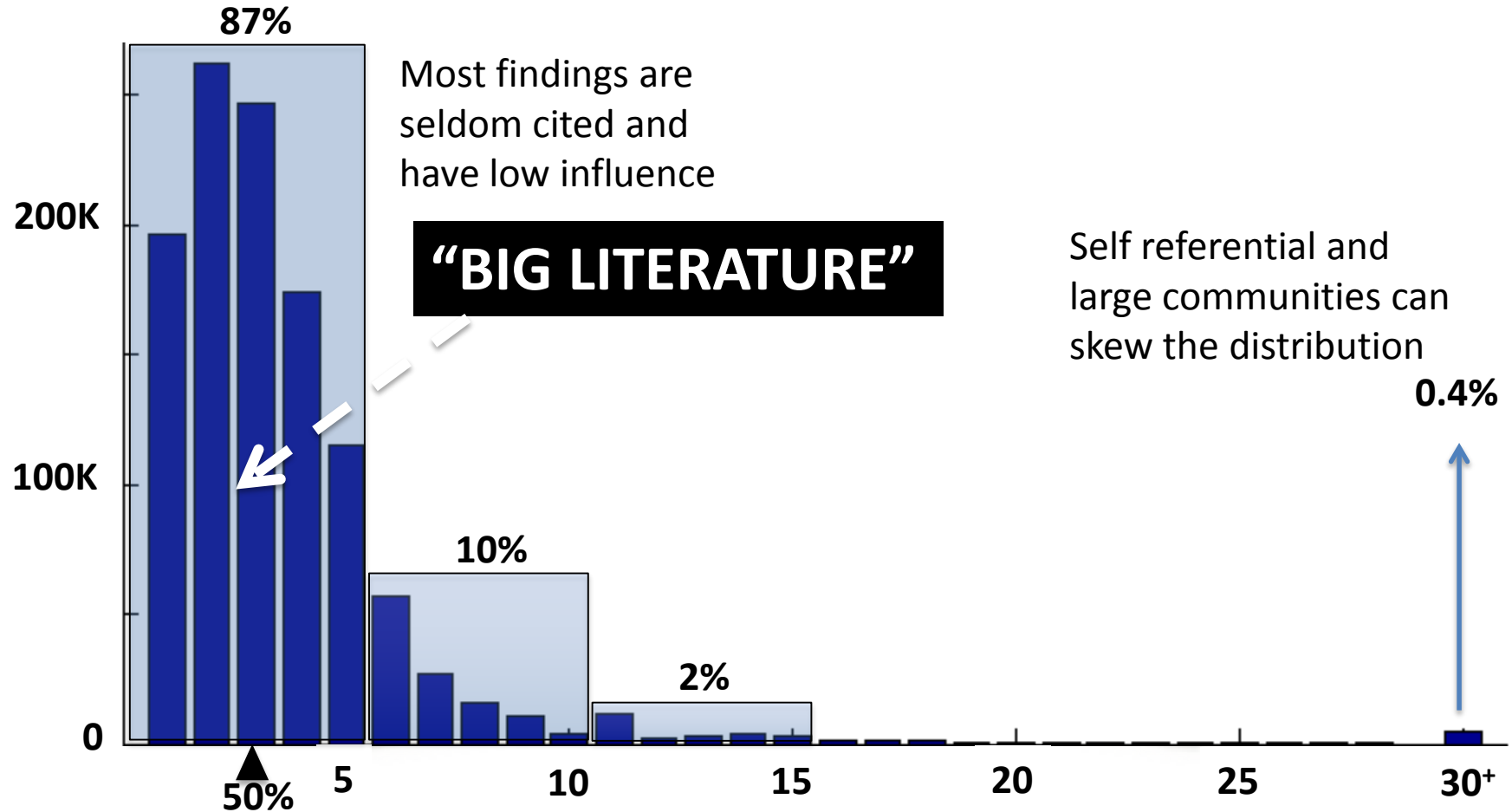
BIG LITERATURE PROBLEM

5-YEAR IMPACT FACTOR OF ISI JOURNAL PUBLICATIONS IN 2012 (1.2 MILLION)



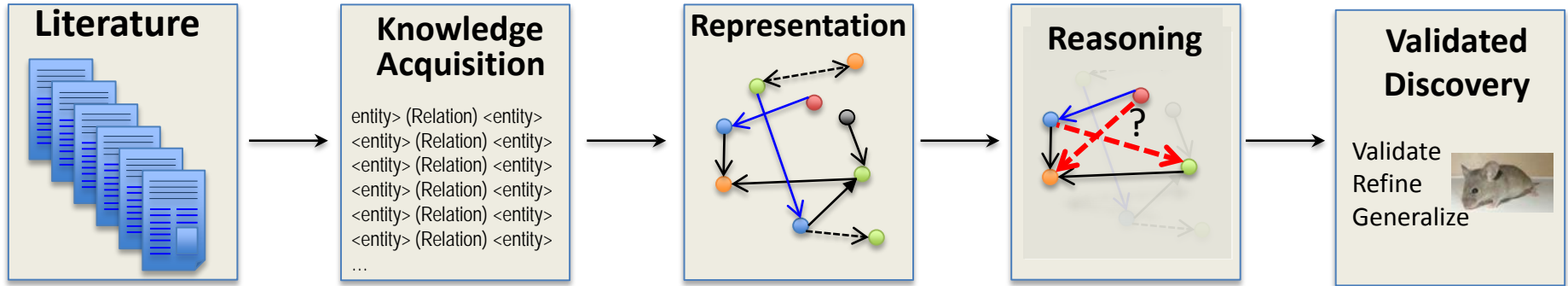
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Too little time to read and to learn

A KNOWLEDGE INTEGRATION TOOLKIT: *KnIT*



KNOWLEDGE ACQUISITION

- Survey relevant text
- Extracts relevant entities (human proteins called kinases)
- Model each entities as a points in feature space: these features are coordinates that form an aggregate “text signature” of the entity

KNOWLEDGE REPRESENTATION

- A graph represents similarity relationship among entities.
- Helps visualize hidden literature connections between entities.
- Coloring may reveal sub-graphs of clusters of interest.
- Critically, a sub-graph may contain unexpected entities
- Hypothesis: perhaps these unexpected entities share their neighbors’ functional properties

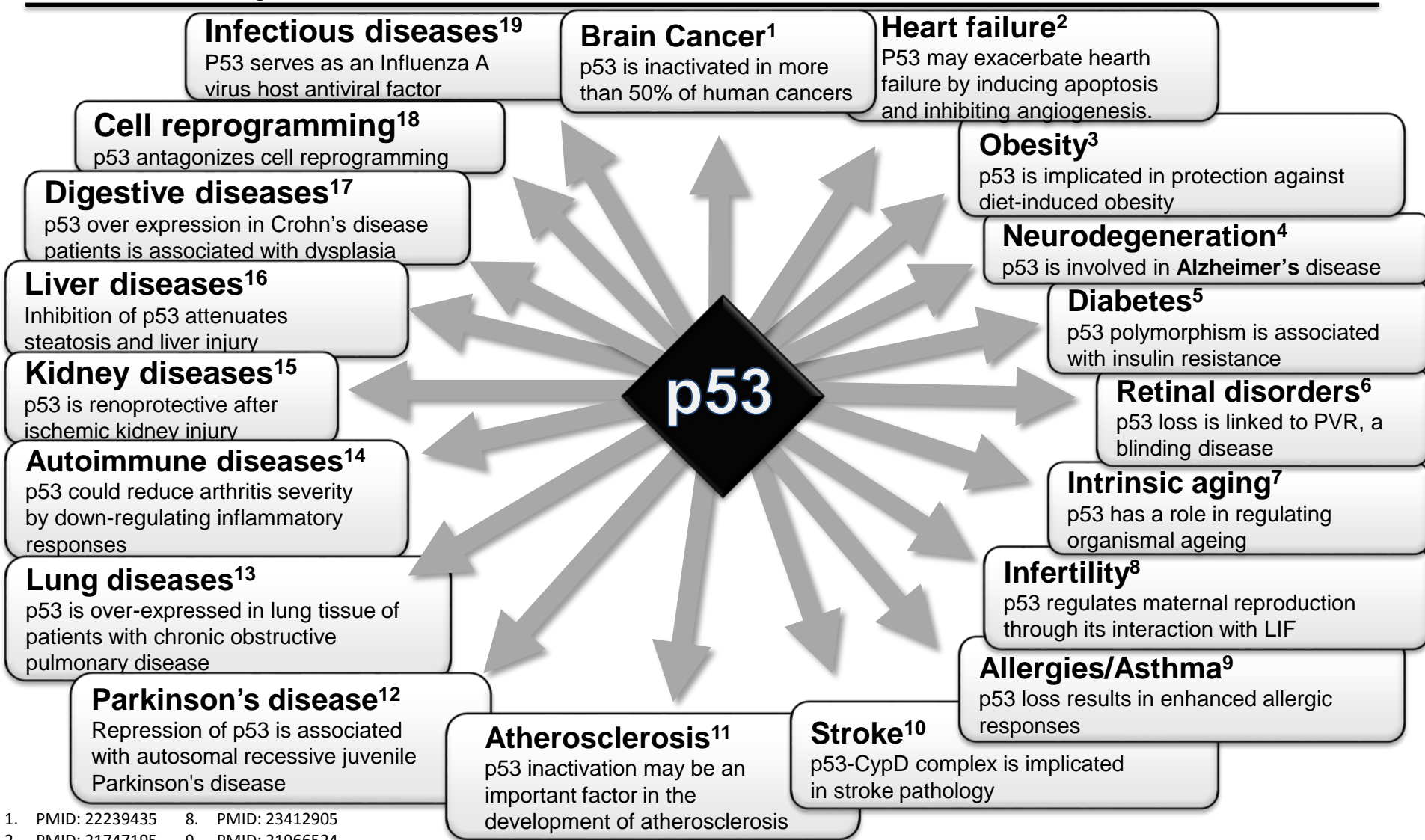
REASONING

- Diffuses information among linked entities
- Rank order candidates for further experimentation of novel annotation predictions.
- The domain expert can evaluate the rankings, the supporting evidence, and choose which to pursue experimentally.

*To accelerate scientific progress
by integrating mining, visualization, and analytics*

PROOF OF PRINCIPLE

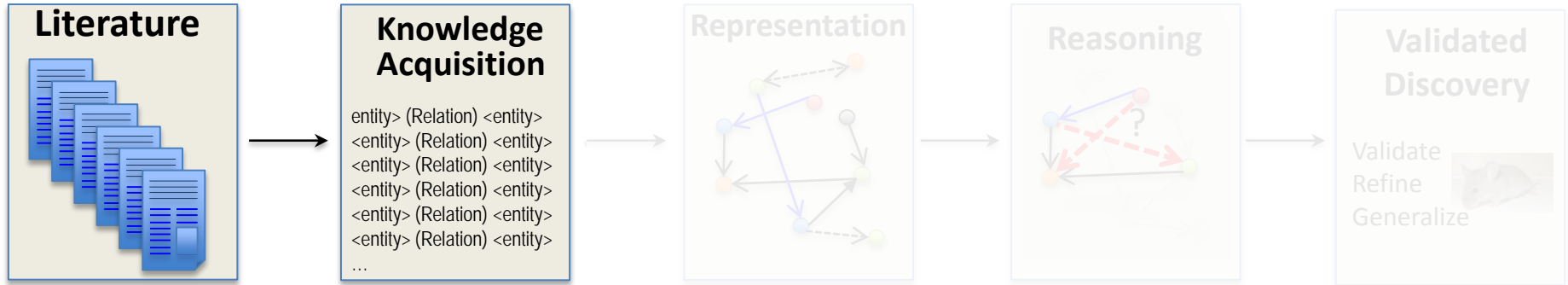
p53 THE “GUARDIAN OF THE GENOME”



- | | | |
|-------------------|--------------------|--------------------|
| 1. PMID: 22239435 | 8. PMID: 23412905 | |
| 2. PMID: 21747195 | 9. PMID: 21966524 | |
| 3. PMID: 23412343 | 10. PMID: 22726440 | 15. PMID: 23222126 |
| 4. PMID: 22387179 | 11. PMID: 10086392 | 16. PMID: 23211317 |
| 5. PMID: 23269546 | 12. PMID: 19801972 | 17. PMID: 17676397 |
| 6. PMID: 22901751 | 13. PMID: 20423464 | 18. PMID: 19668186 |
| 7. PMID: 11780111 | 14. PMID: 18341615 | 19. PMID: 22105999 |

p53 is the “first responder” to cellular stress, as a result it is linked to a plethora of brain diseases

A KNOWLEDGE INTEGRATION TOOLKIT: *KnIT*



KNOWLEDGE ACQUISITION

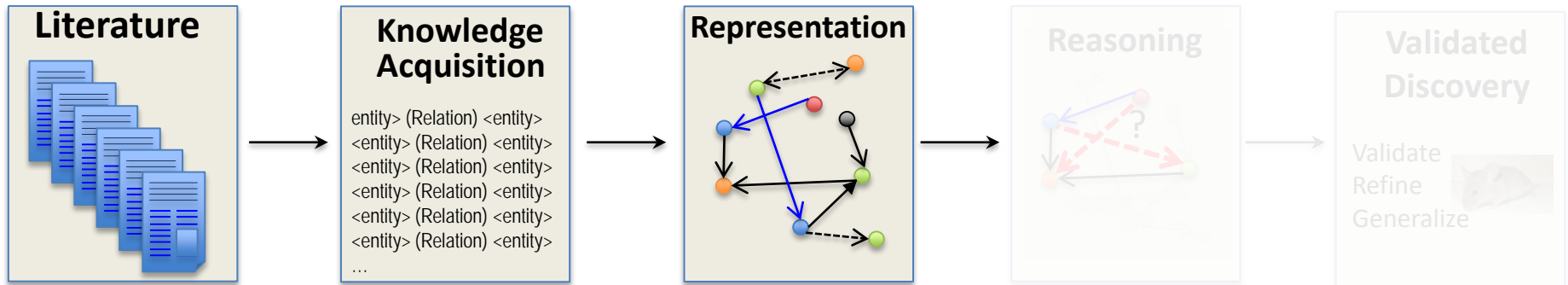
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Biological Entity Similarity Network



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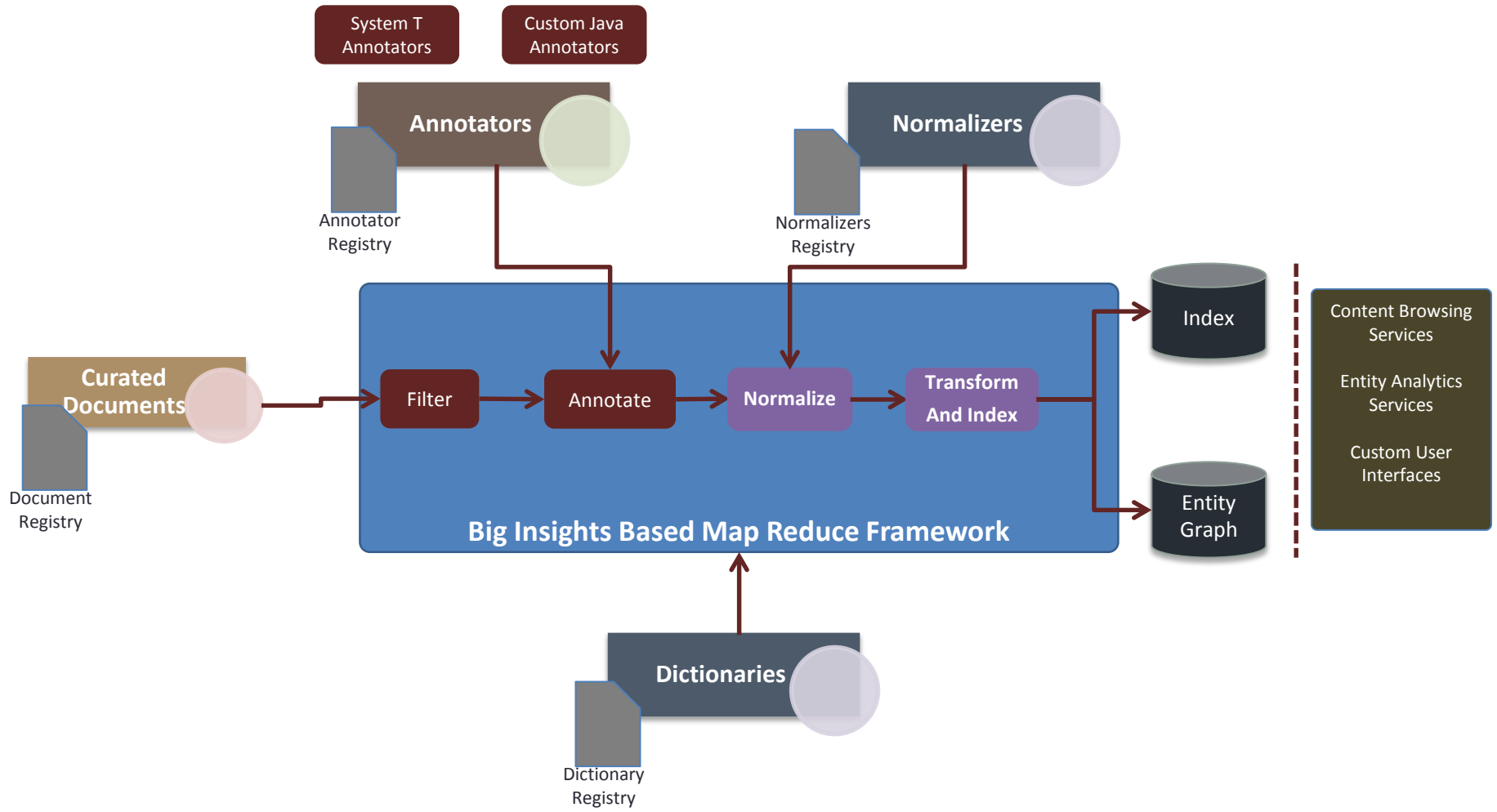
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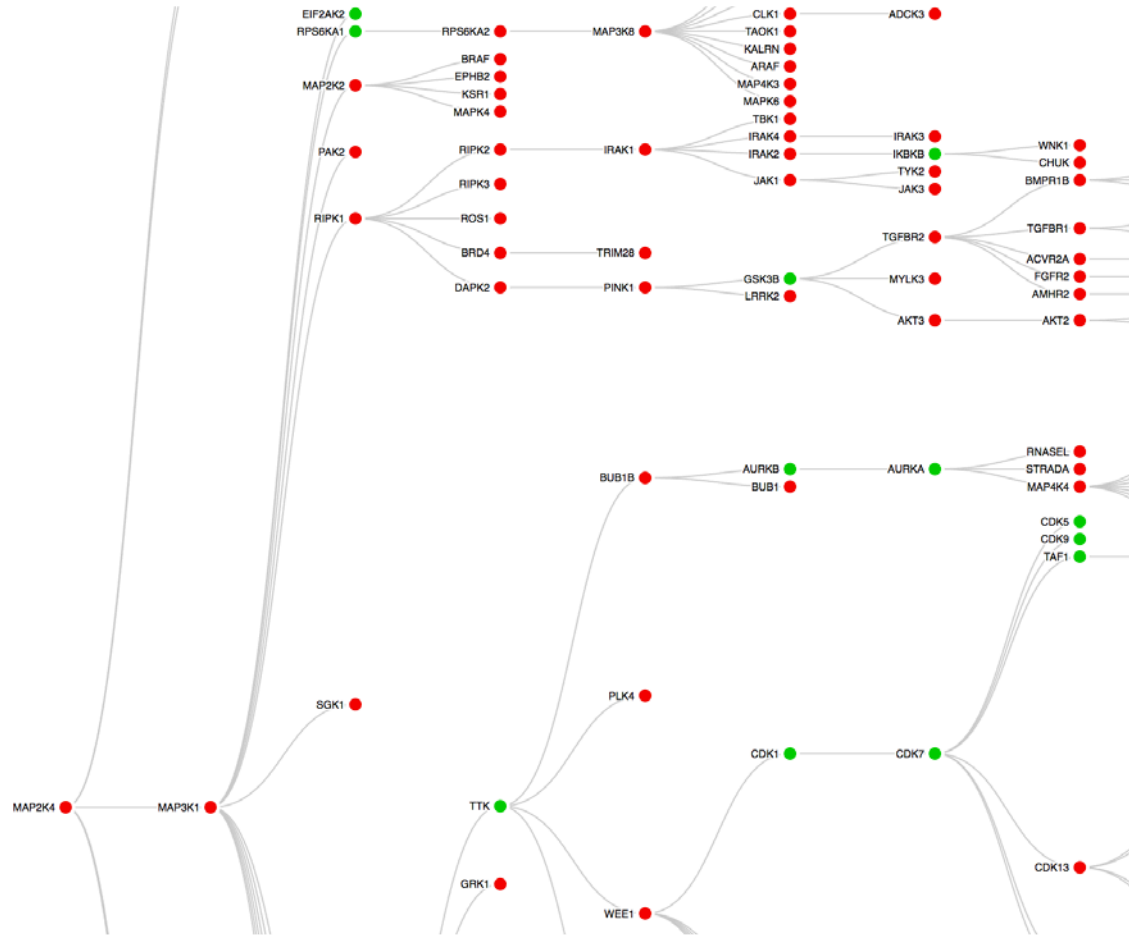
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DISCOVERY END TO END WORKFLOW



Kinase Similarity Tree



Algorithm

Algorithm 1 *Create an n-ary similarity tree from a set of entities*

Input: entities, n

Output: n-ary similarity tree

mostTypicalFV = average(entities)

root = closestTo FV(entities, mostTypical FV)

entities.remove(root)

candidates = {root}

while not entities.isEmpty()

 (e, c) = closestPair(entities, candidates)

 c.addChild(e)

if c.numChildren() == n **then**

 candidates.remove(c)

end if

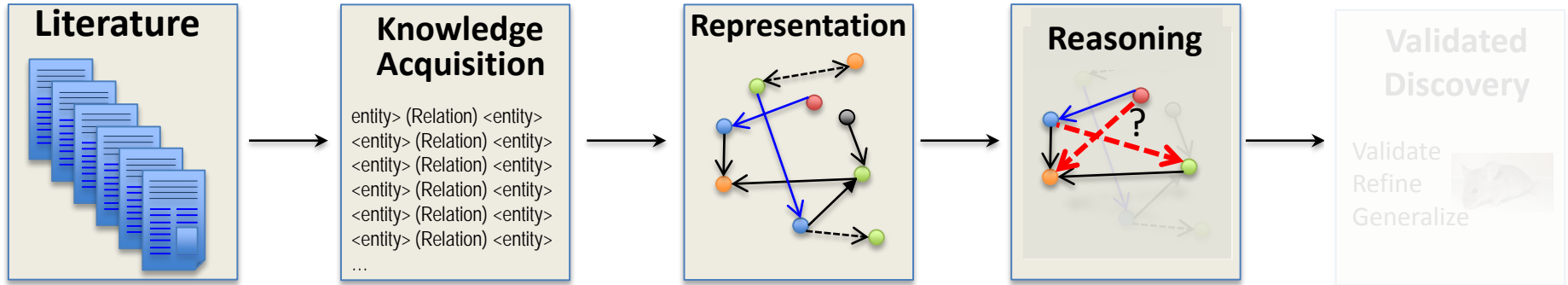
 candidates.add(e)

 entities.remove(e)

end while

Return: root

A KNOWLEDGE INTEGRATION TOOLKIT: *KnIT*



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REASONING

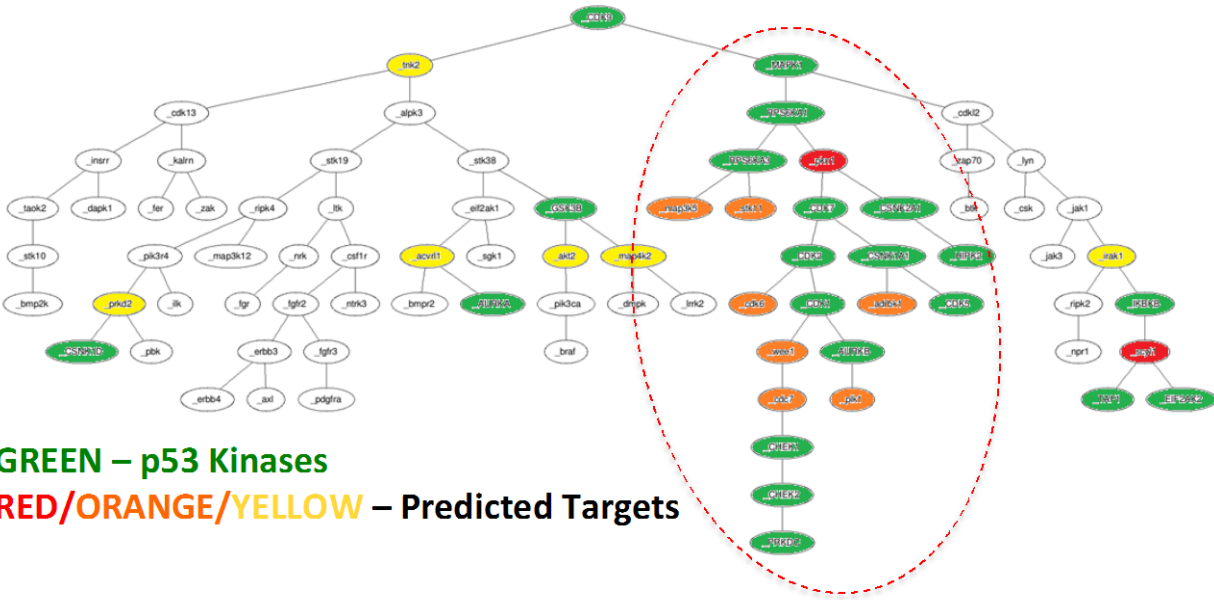
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ENTITY SIMILARITY TREE



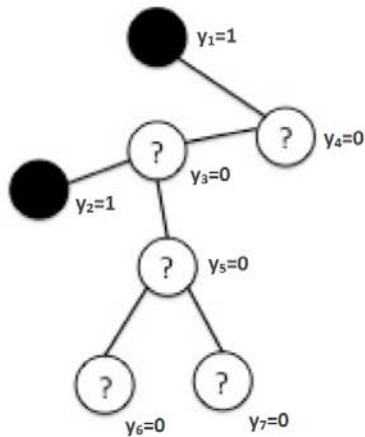
$$F = \begin{bmatrix} f_{11} & f_{12} & \dots & \dots & f_{1n} \\ \vdots & f_{22} & & & \vdots \\ \vdots & & \ddots & & \vdots \\ \vdots & & & \ddots & \vdots \\ f_{n1} & f_{n2} & \dots & \dots & f_{nn} \end{bmatrix}$$



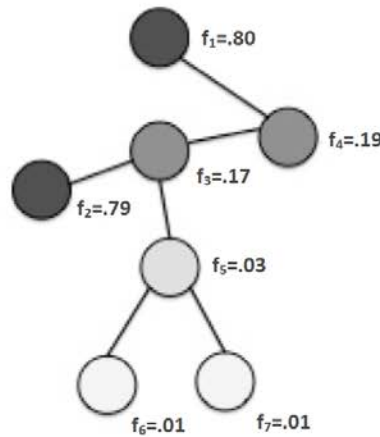
RETROSPECTIVE CONTROL

A.

Label p53 Kinases

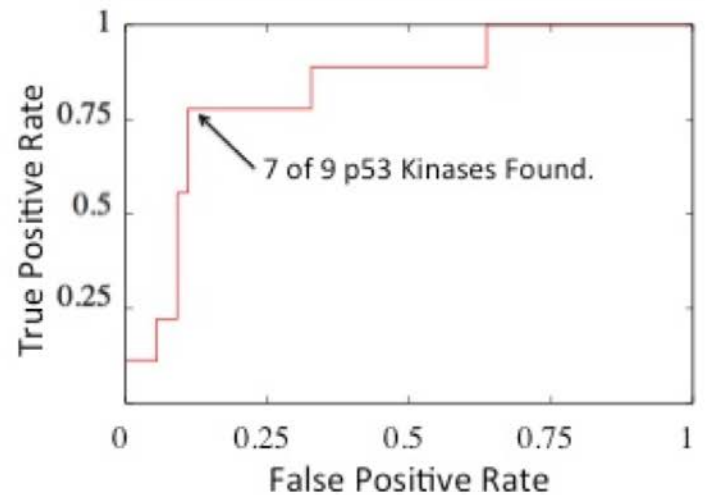


Diffuse Labels



B.

P53 KINASES DISCOVERED POST-2003
BASED ON THE PRE-2003 LITERATURE

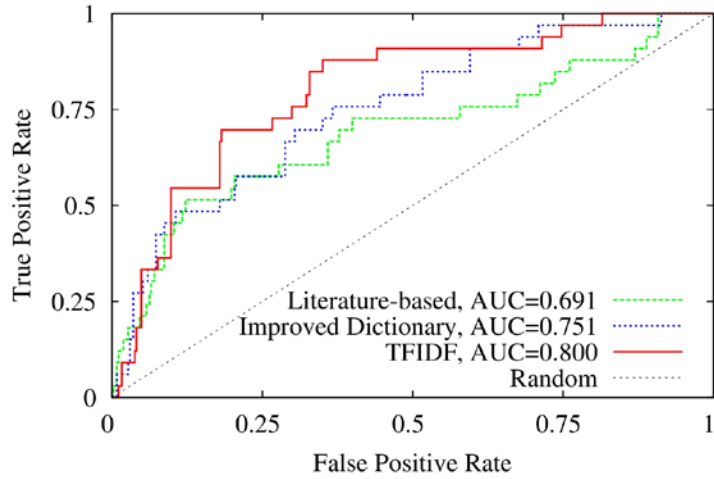


**GRAPH INFORMATION DIFFUSION OF P53 KINASE LABELS KNOWN PRIOR TO 2003
RECOVERS P53 KINASES DISCOVERED AFTER 2003**

Example of this type of diffusion: Lisewski et al *CELL* (2014)

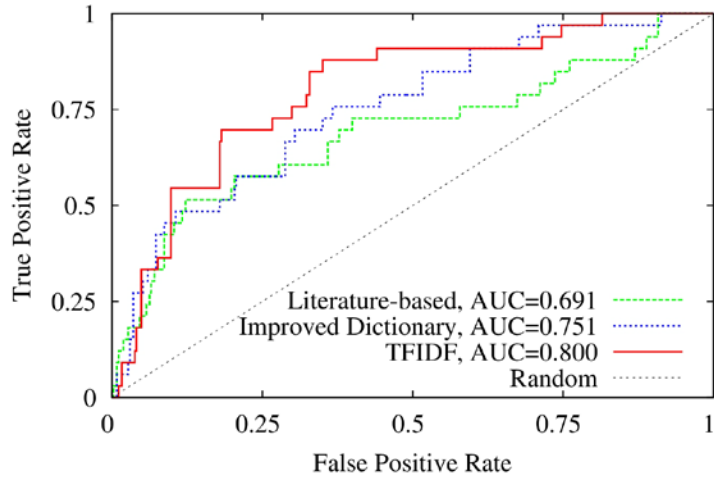
LEAVE-ONE-OUT EXPERIMENTS ALSO SUGGEST THAT THIS APPROACH CAN PREDICT WHICH KINASES TARGET A GIVEN PROTEIN

RECOVERY OF KNOWN P53 KINASES

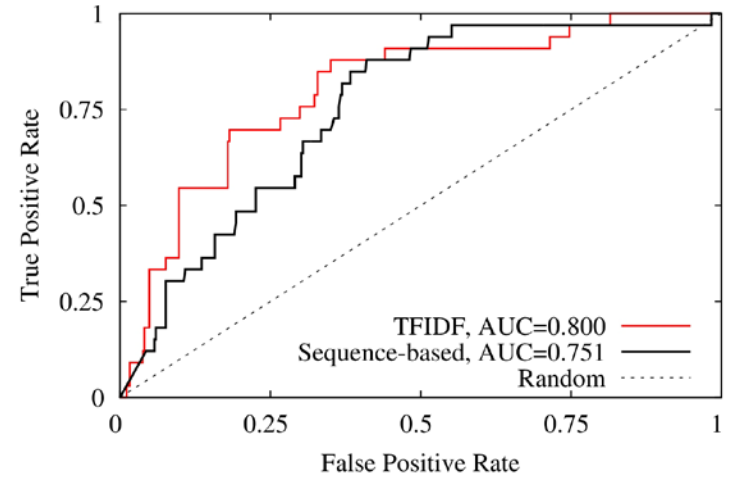


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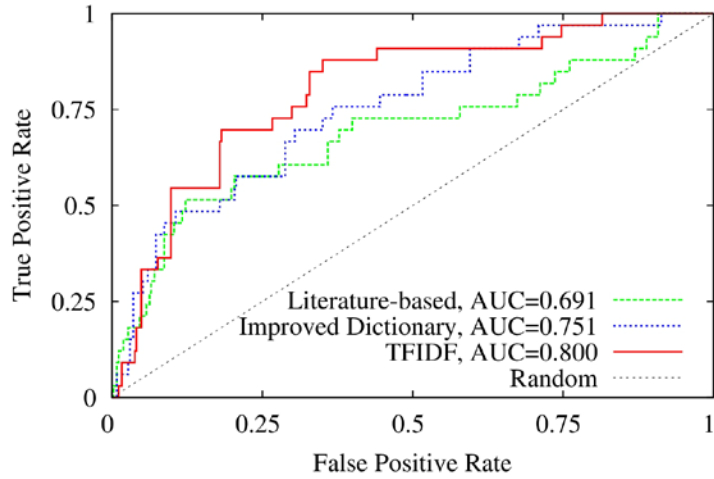


COMPARISON TO SEQUENCE ANALYSIS

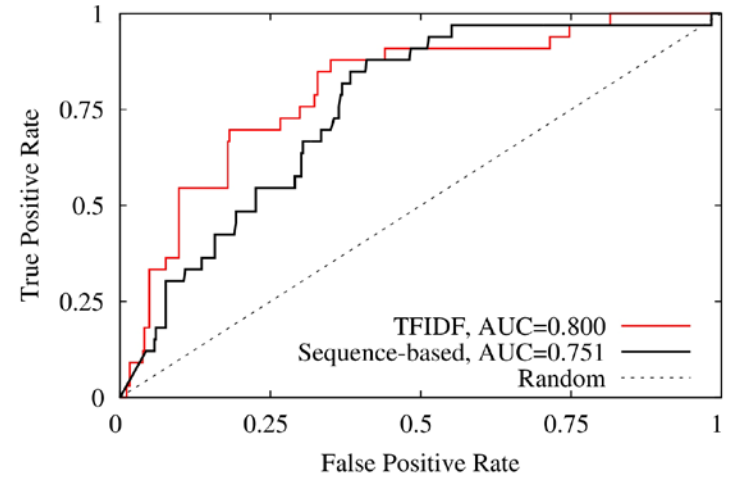


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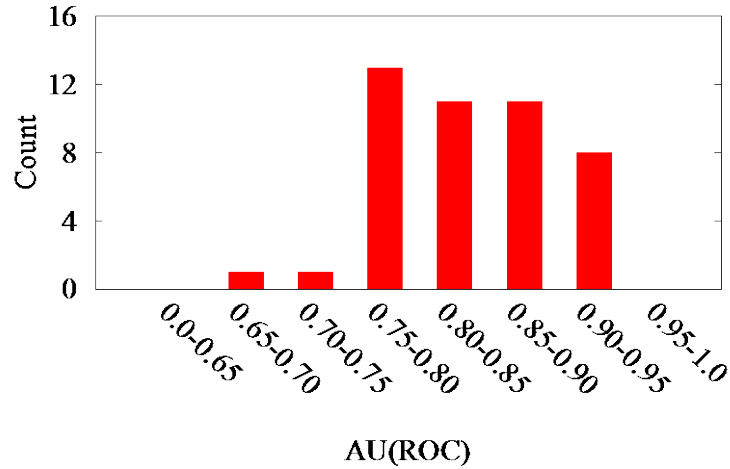
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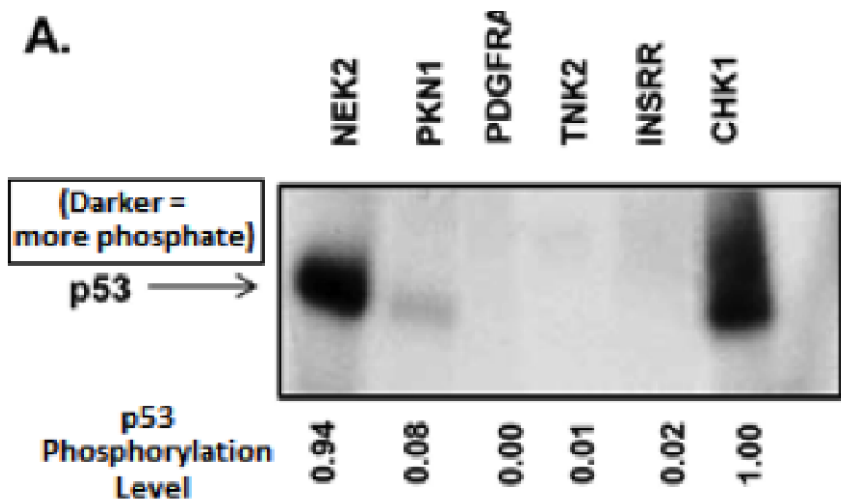


KINASE PREDICTIONS IN 45 OTHER PROTEINS



BONA FIDE EXPERIMENTAL VALIDATION

A.



B.

CONDITIONS

(Darker = protein is present)

p53 Antibody Negative Control Antibody

NEK2 Present? + - + -

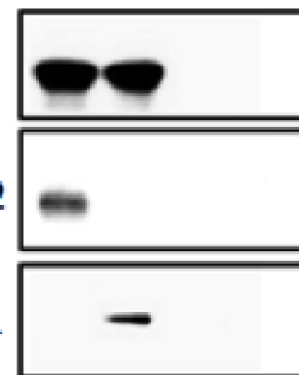
PKN1 Present? - + - +

RESULTS

Detecting p53

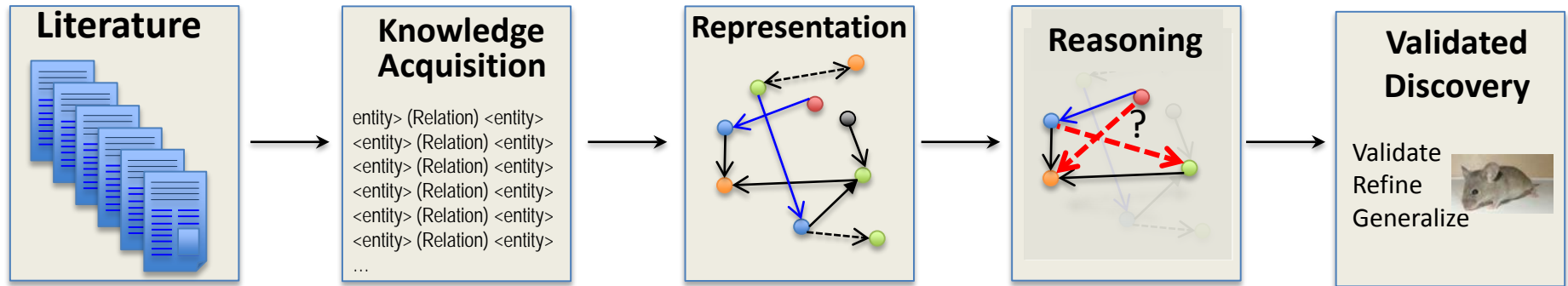
Detecting NEK2

Detecting PKN1



Lane 1 2 3 4

A KNOWLEDGE INTEGRATION TOOLKIT: *KnIT*



- Laboratory support for p53 kinases predicted from text mining
- Proof of principle for a strategy to predict some unknown fact from the scientific literature
- A first step to predict new connections based on everything else that is known.
- Future: more work needed to
 - Broaden the scope of proteins and functions
 - Comprehensive networks of interactions
 - To gather a more complete understanding of the mechanisms behind disease
 - Translate this into clinical impact.
 - Test this approach of mining literature to identify hidden relationships beyond cancer and beyond biology to other areas of human though where text is a bottleneck.
- Such acceleration of discovery is not only desirable, but also indispensable for human flourishing.