



Pattern Recognition and Anomaly Detection in Bookkeeping Data

Asian Bureau of Finance and Economic Research (ABFER)

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Carnegie Mellon University



Coauthors Influences

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Outline

Use Case 1: Use Graph Mining Tools to Detect Journal entry anomalies

Broad Lessons

Use Case 2: Use Graph Mining Tools to Quantify Account Classification and Aggregation

Why this is cool?

USE CASE:

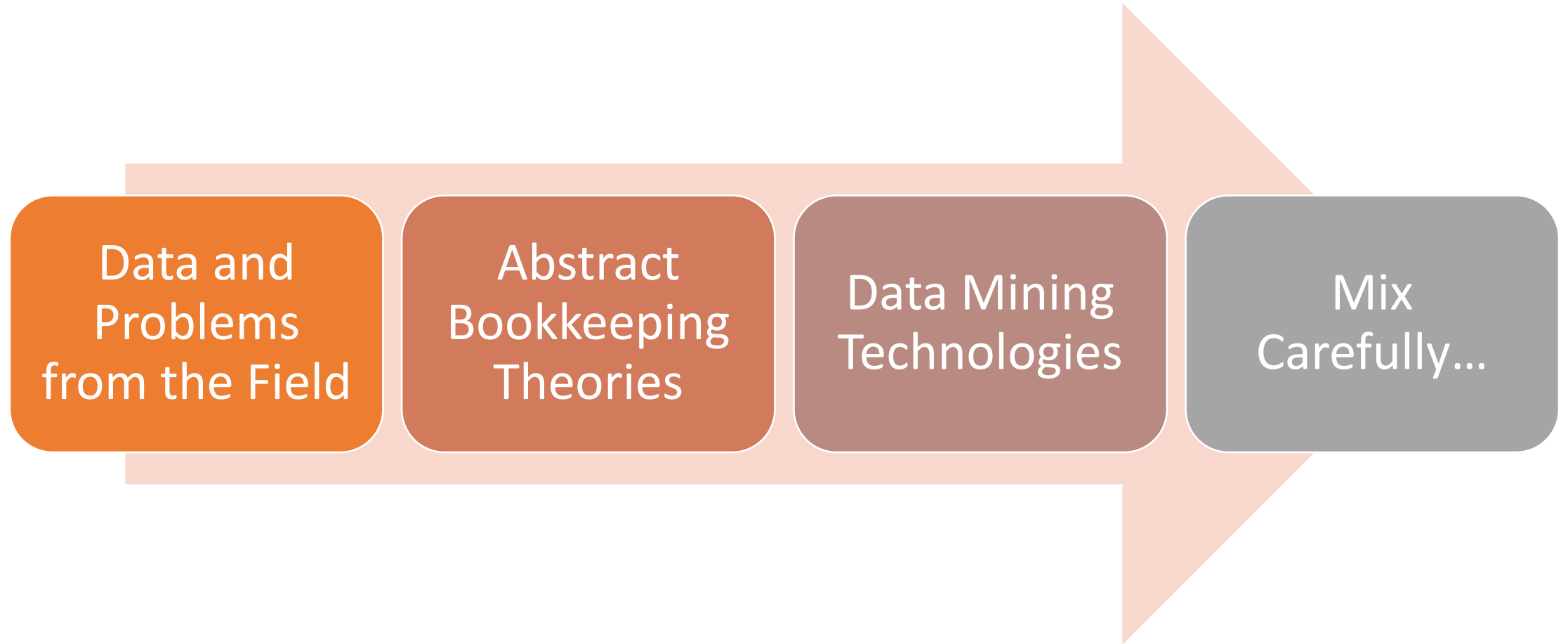
Using Graph Mining to Detect Anomalies in Bookkeeping Data

Principal Investigators:

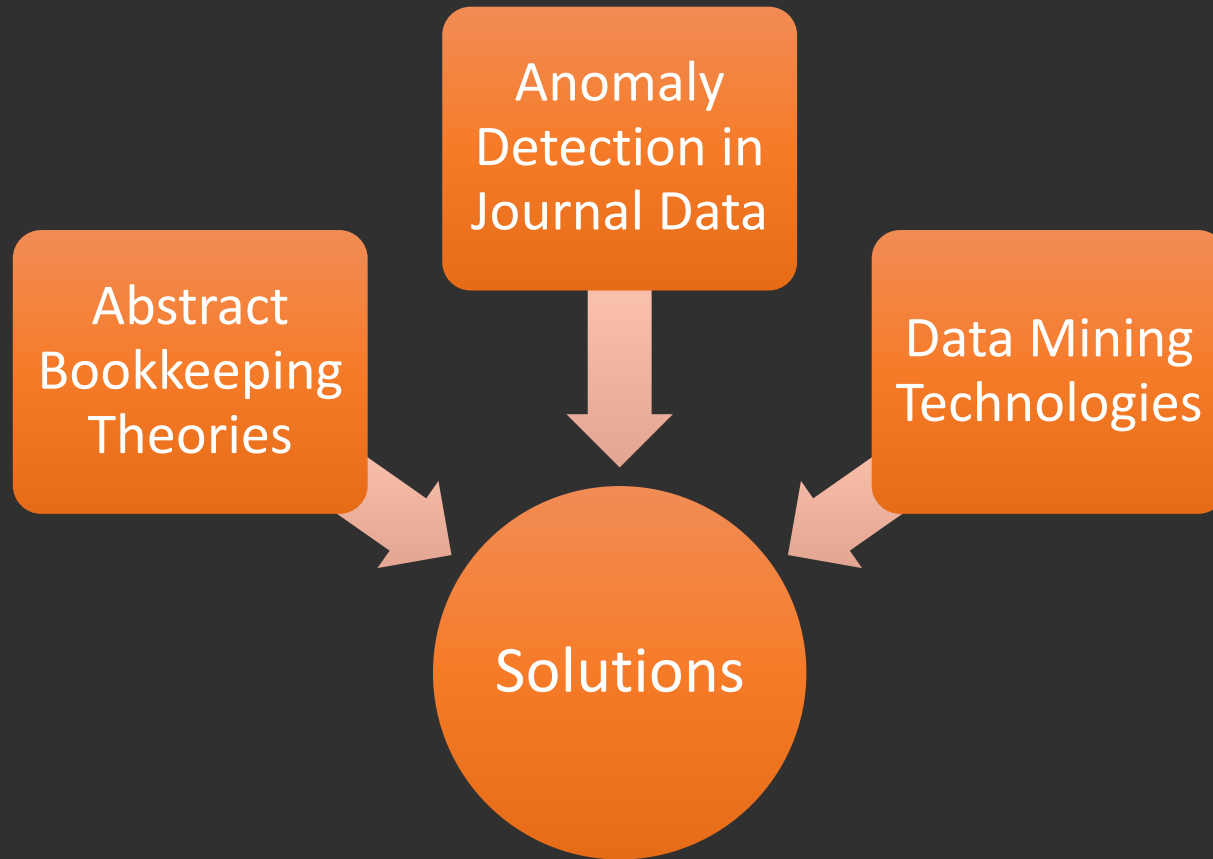
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University

Three Ingredients



Ingredient 1: Data and Problems from Field



Journals and Ledgers

General Journal				
Year	Transaction Details	GL Ref	D	C
Month				
Dec 15	Cash	101	12000	
	Share Capital	301		12000
	Establishment of XYZ			
Dec 17	Land	201	15000	
	Accounts Payable	401		15000
	Buildings	202	10000	
	Cash	101		10000
	Acquisition of fixed assets			

Land Acc. No. 201				
Year	Transaction Details	D	C	Balance
Month				
Dec 17	Acquisition of fixed assets	15000		15000

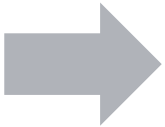
Accounts Payable Acc. No. 401				
Year	Transaction Details	D	C	Balance
Month				
Dec 17	Acquisition of fixed assets		15000	(15000)

Buildings Acc. No. 202				
Year	Transaction Details	D	C	Balance
Month				
Dec 17	Acquisition of fixed assets	10000		10000

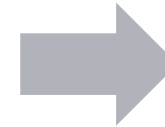
Cash Acc. No. 101				
Year	Transaction Details	D	C	Balance
Month				
Dec 15	Establishment of XYZ	12000		12000
Dec 17	Acquisition of fixed assets		10000	2000

Our Project GOAL

**Journal
Entries
Data**

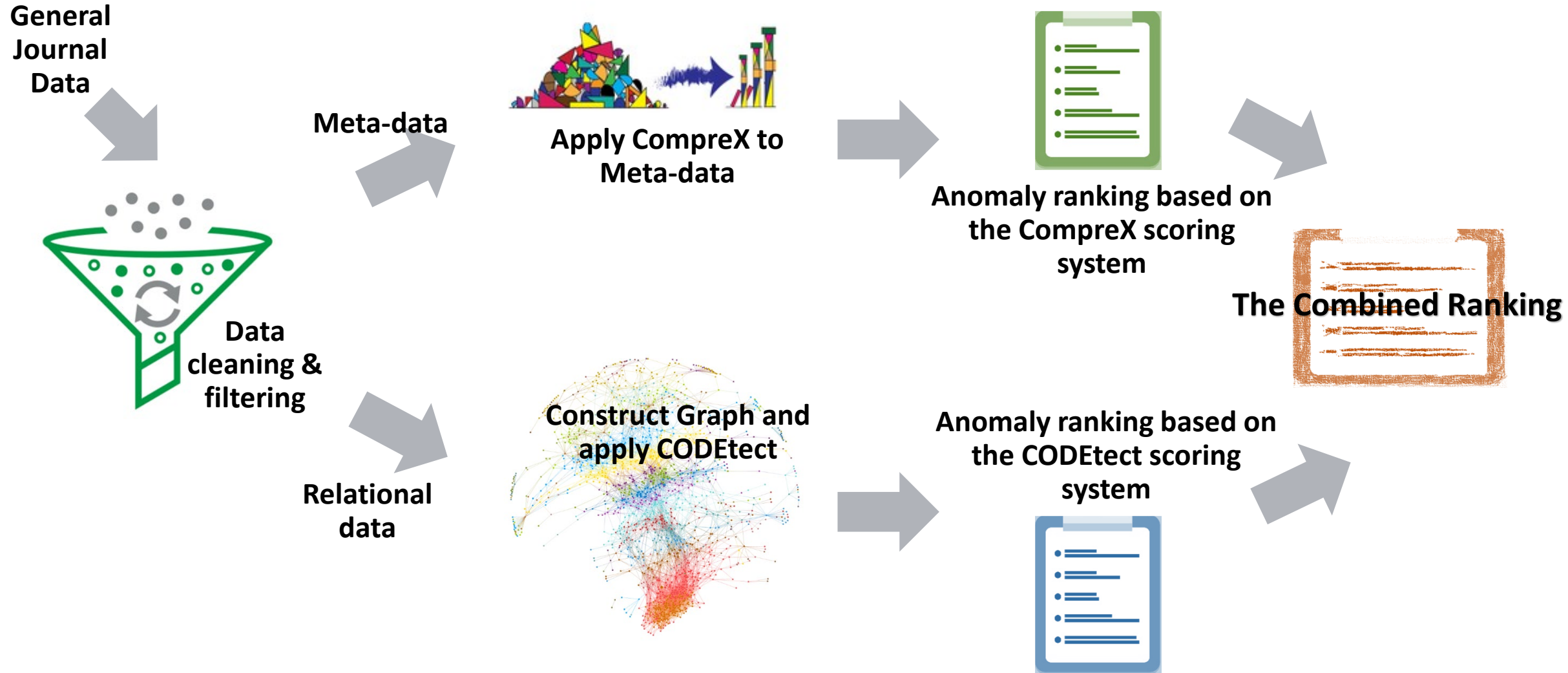


**general, scalable, explainable, and
unsupervised detection model
leveraging meta & relational data**

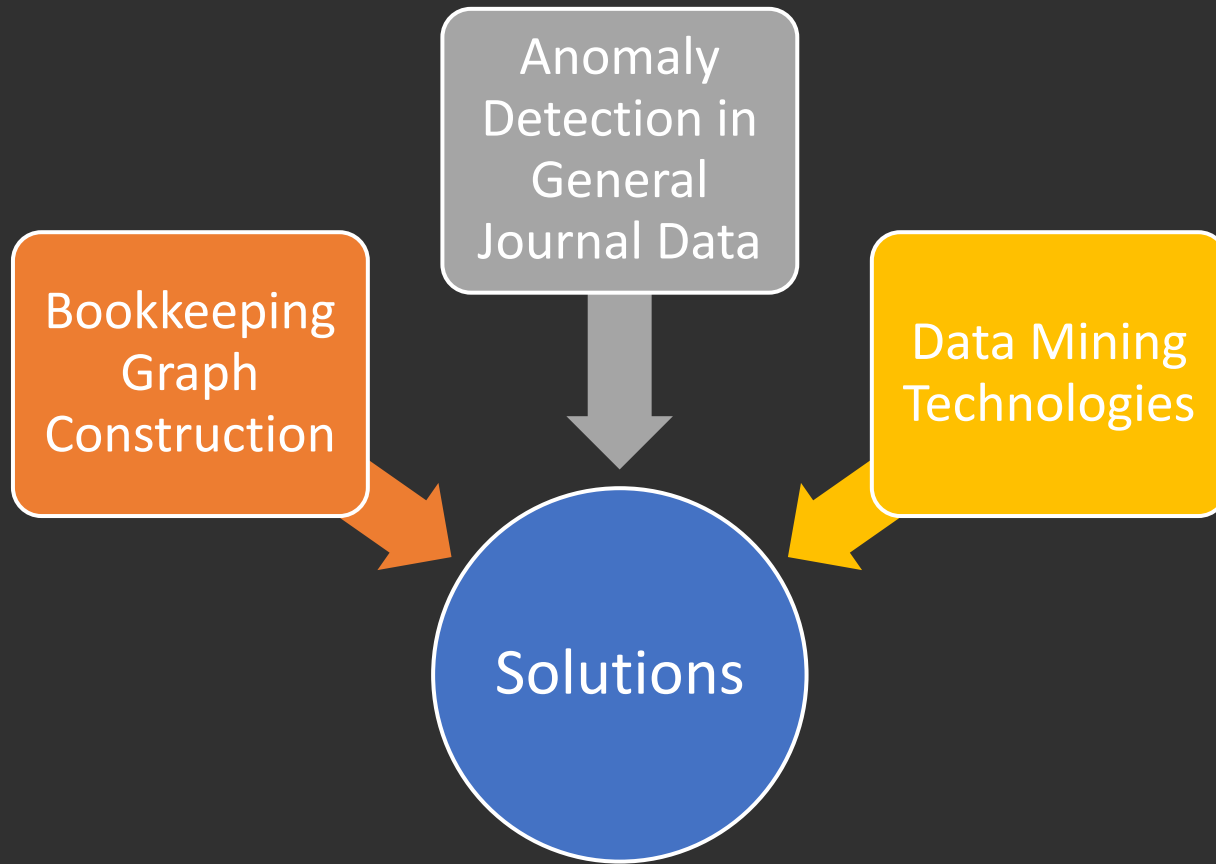


**Journal
Entries
Anomaly
Ranking**

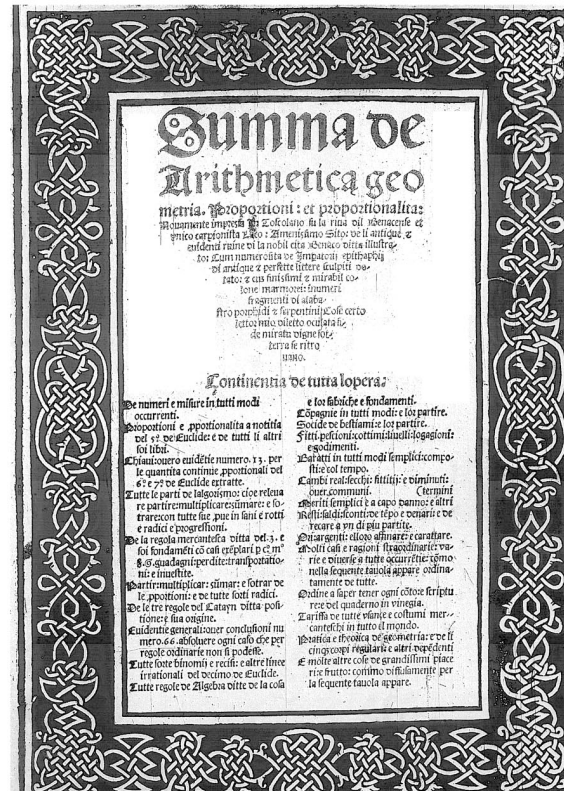
Solutions



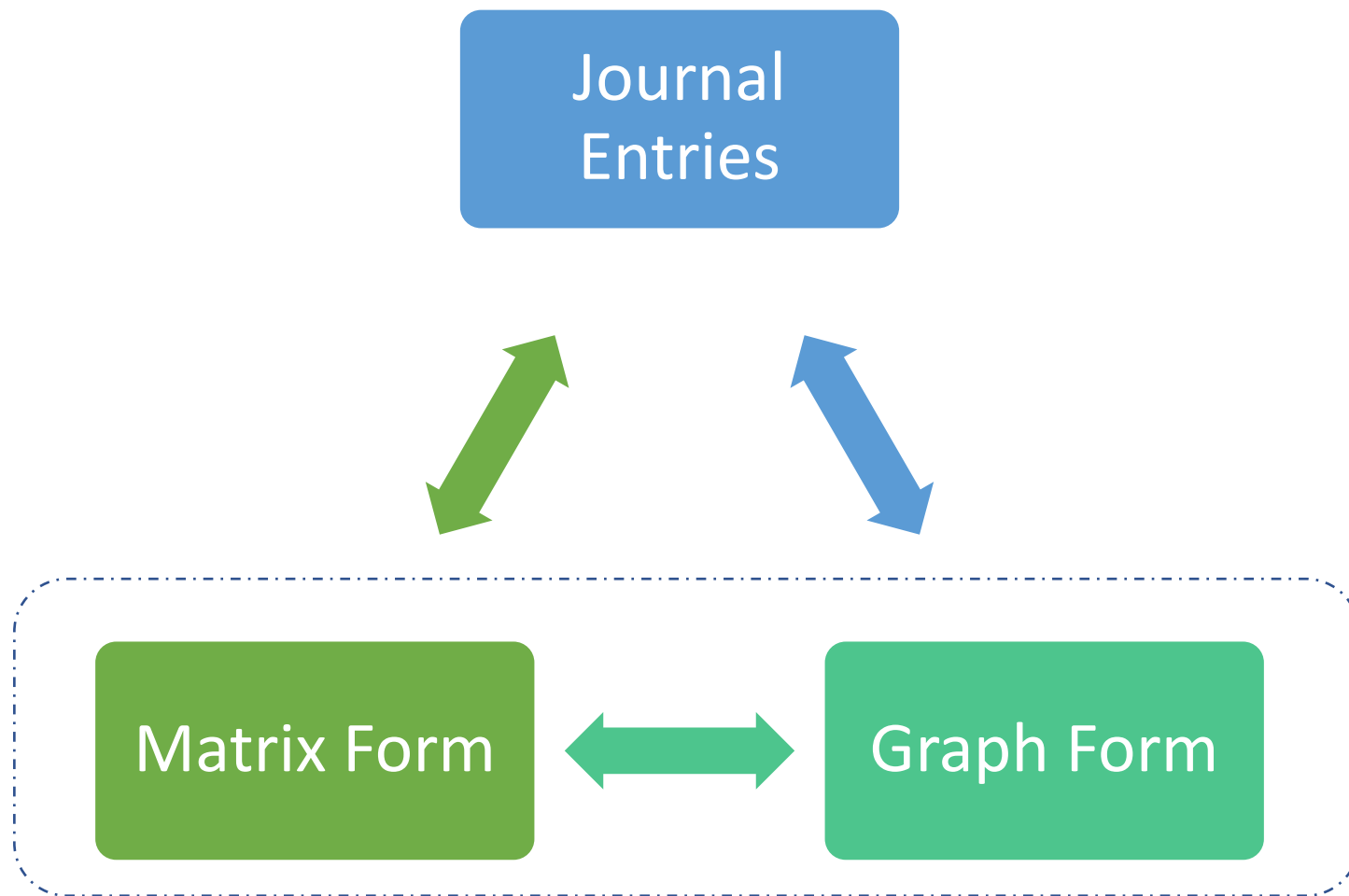
Ingredient 2: Abstract Bookkeeping Theories



Luca Pacioli: Summa de Arithmetica ... (1494 AD)



Double Entry Bookkeeping's "Trifecta"

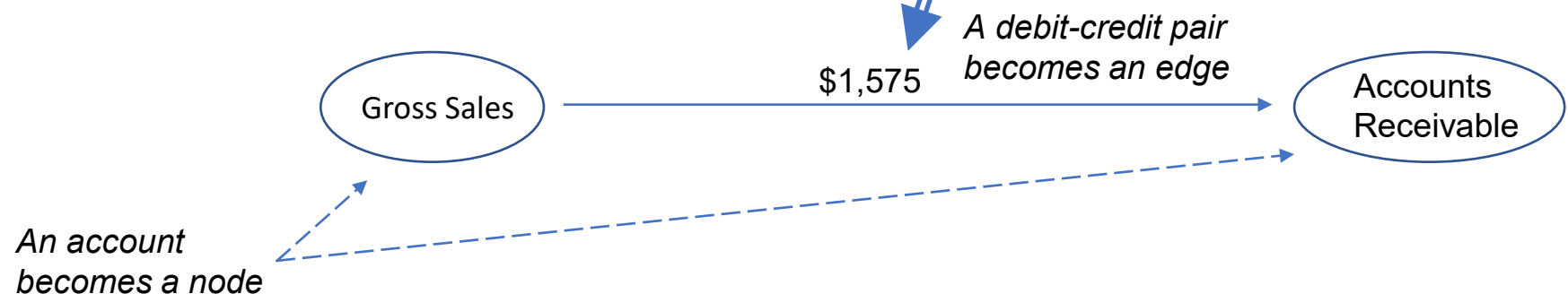


"Double-entry bookkeeping, especially its mathematical structure, has been the author's life-long mystery."
– Yuji Ijiri, 1993

Journal entries can be converted into a DiGraph

- Account => Node
- Entry => Edge
- Debit => Inflow
- Credit => outflow

GL_Account _Number	GL_Journal_ID	GL_User _ID	GL_Reportin g_Amount	CA_FS_Caption	...
40060000 (Revenue)	2016/01- 20160101 - 638621	JJSHEN	-1575	Gross Sales (GSL)	...
10415000 (Assets)	2016/01- 20160101 - 638621	JJSHEN	1575	Accounts Receivable (ARV)	...
...

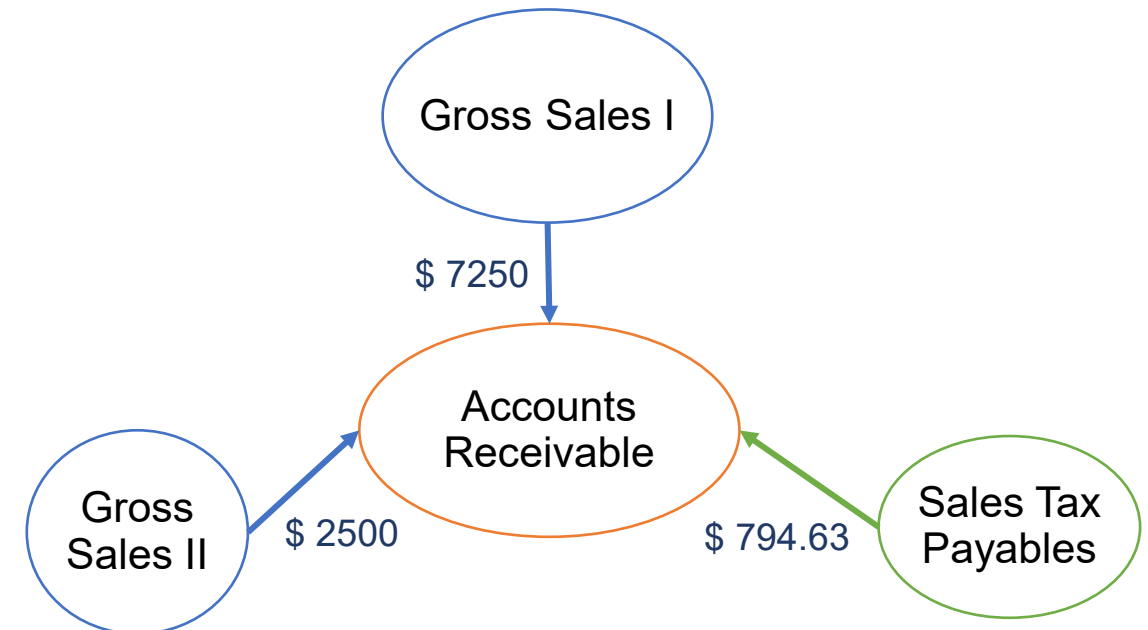


Journal entries can be converted into a DiGraph

HW 2016/01-20160101 -639167

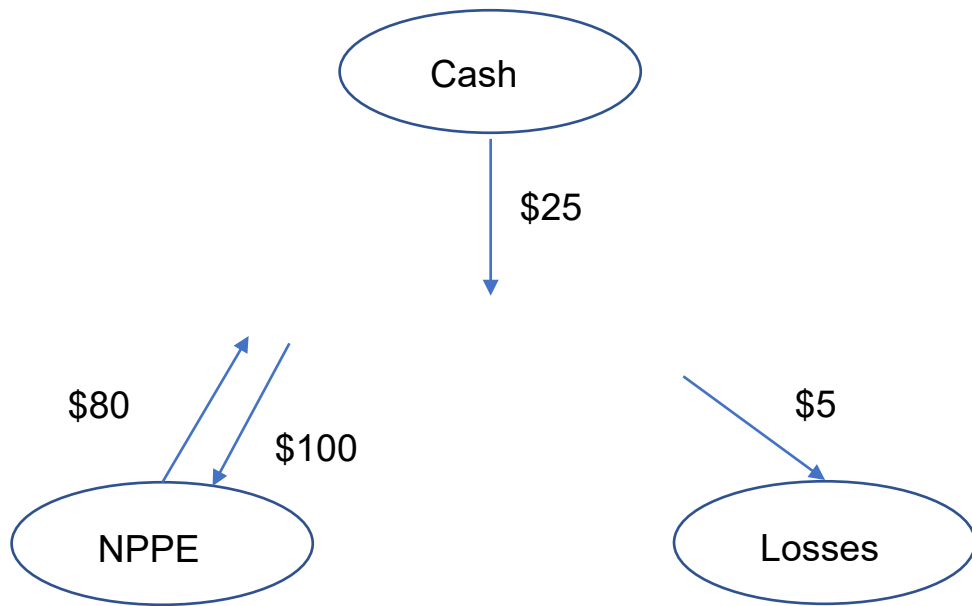
GL_Account Number	CA_FS_Caption	Cr/Dr	GL_Reporting Amount
40020000 (Revenue)	Gross Sales (GSL)	C	-7250
40020000 (Revenue)	Gross Sales (GSL)	C	-2500
20830000 (Liabilities)	Sales Tax Payables (STP)	C	-794.63
10390000 (Assets)	Accounts Receivable (ARV)	D	10544.63

Graph Construction Solution

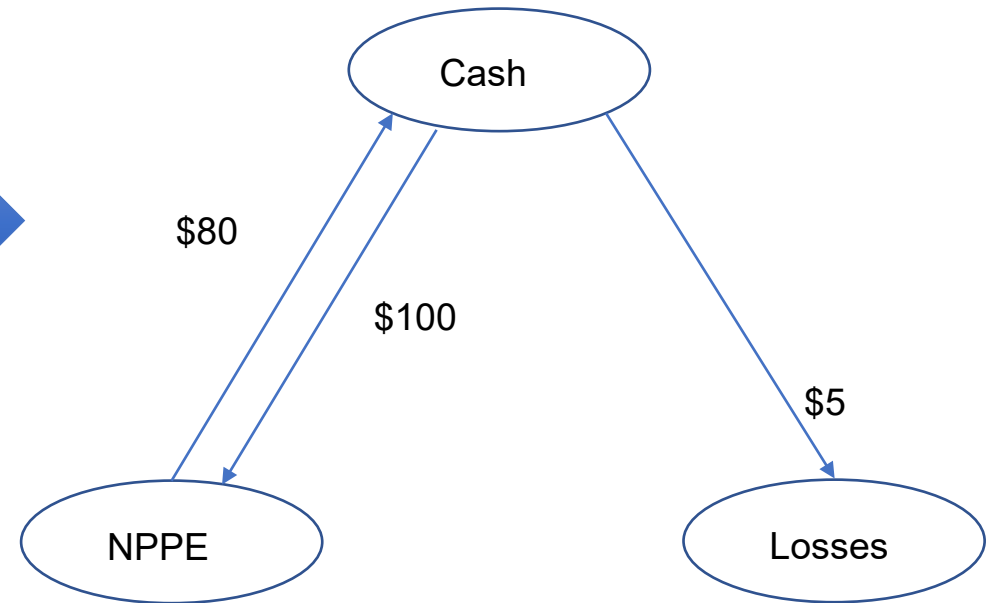


Journal entries can be converted into a DiGraph

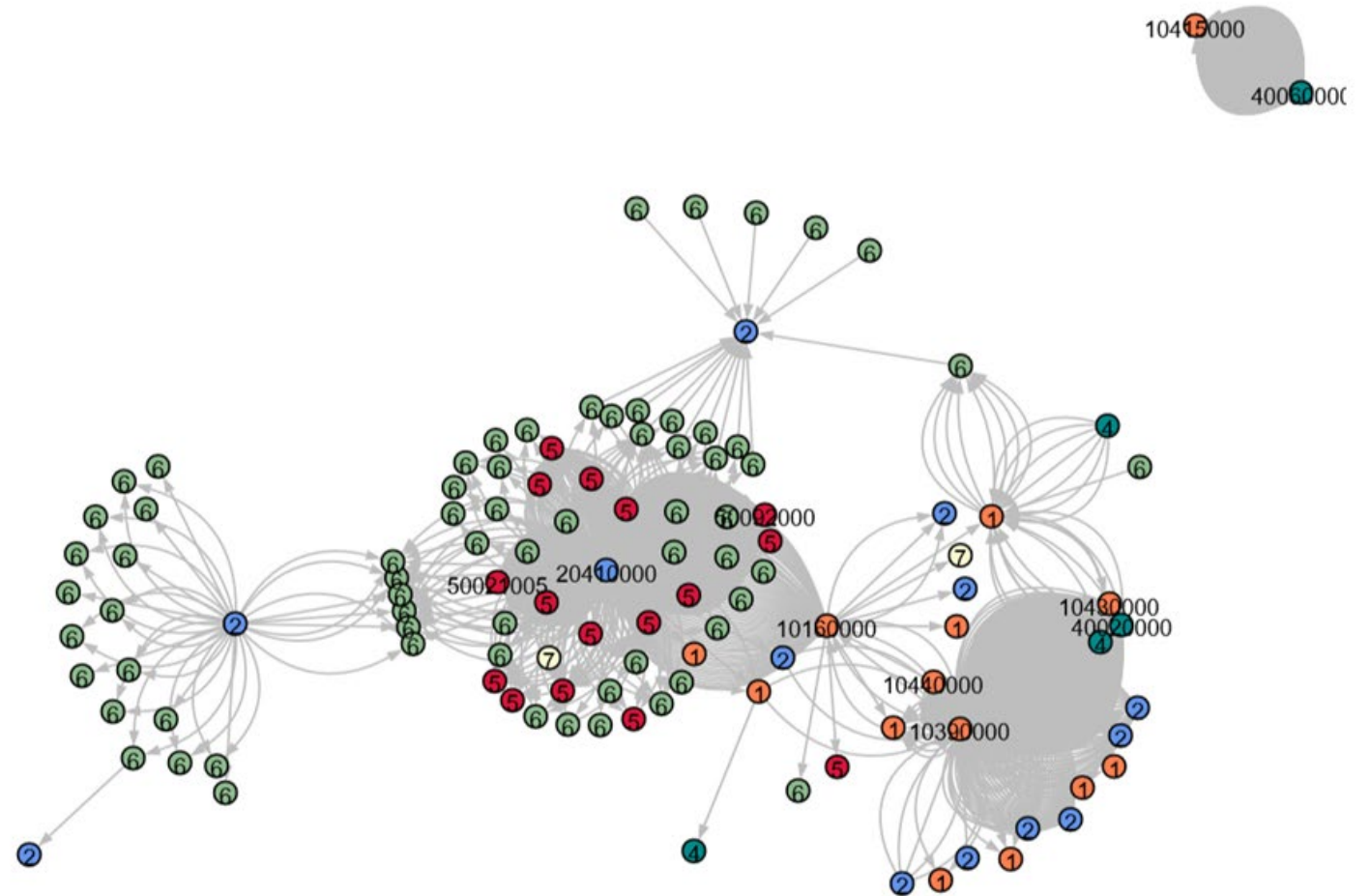
Compound Journal Entries



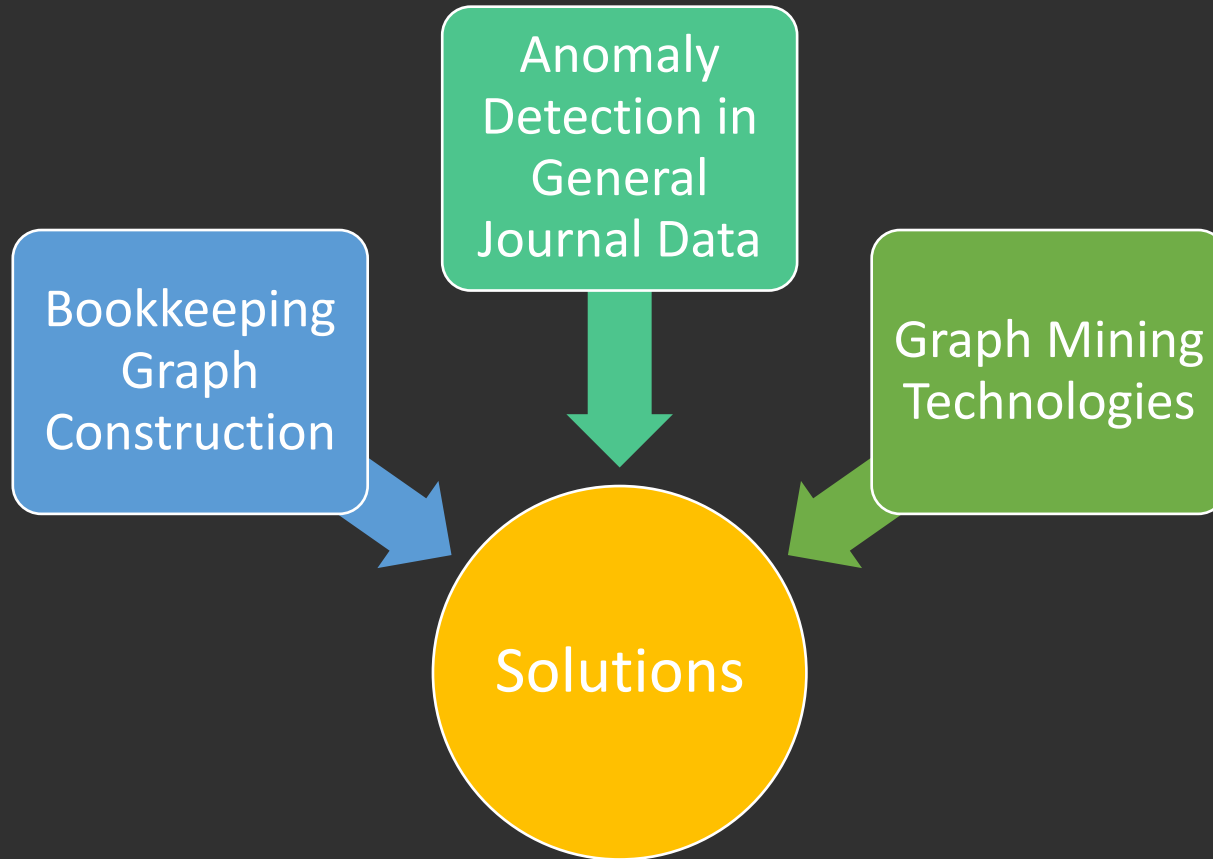
Graph Construction Solution



Bookkeeping Graph for a 10- day Window

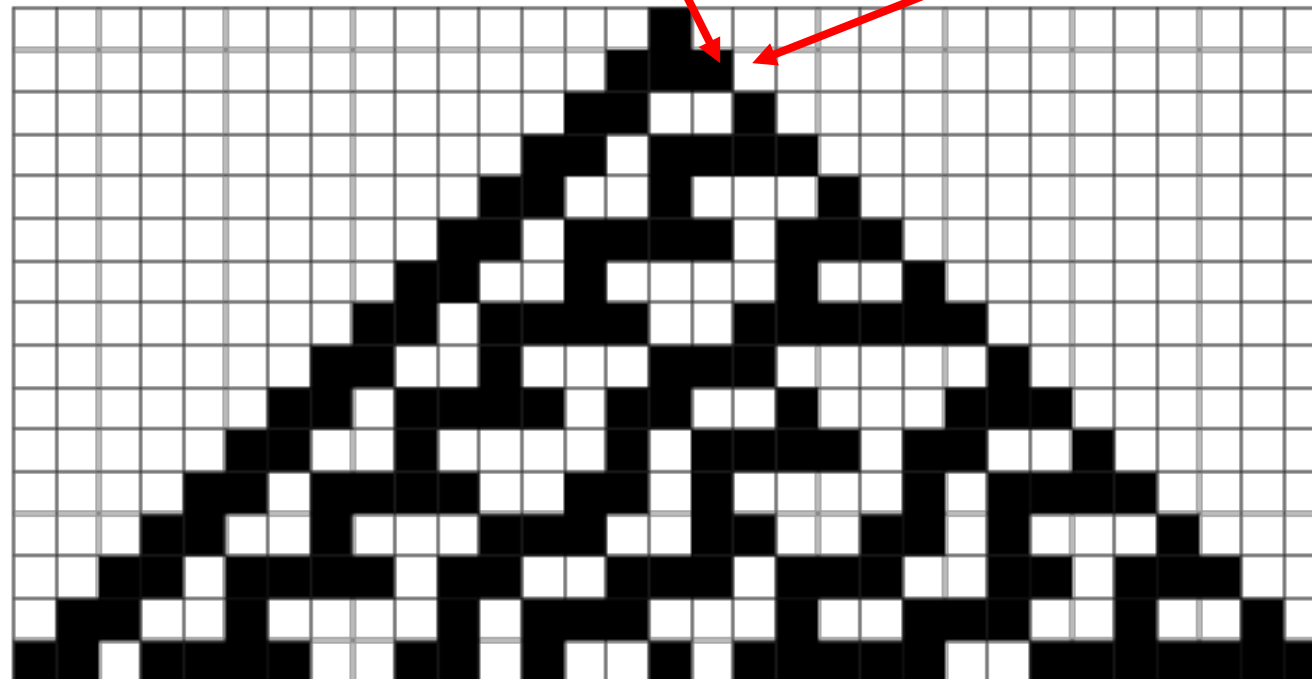
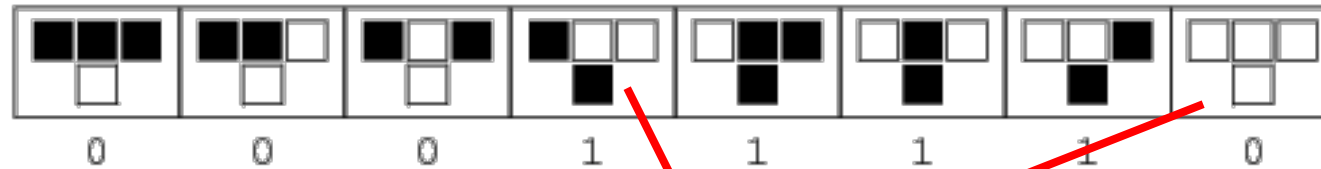


Ingredient 3: Data Mining Technologies

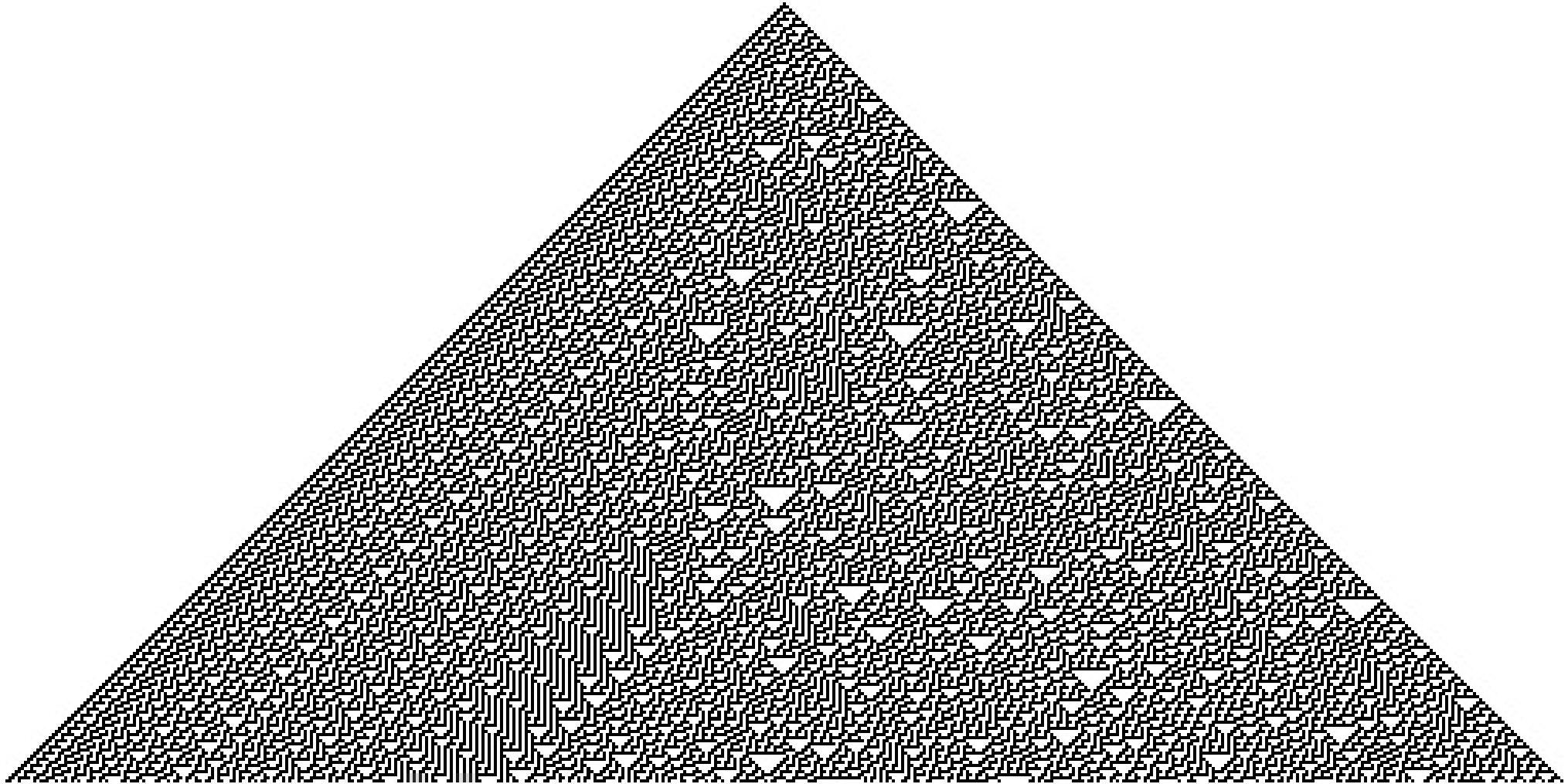


Stephen Wolfram's ECA Rule-30

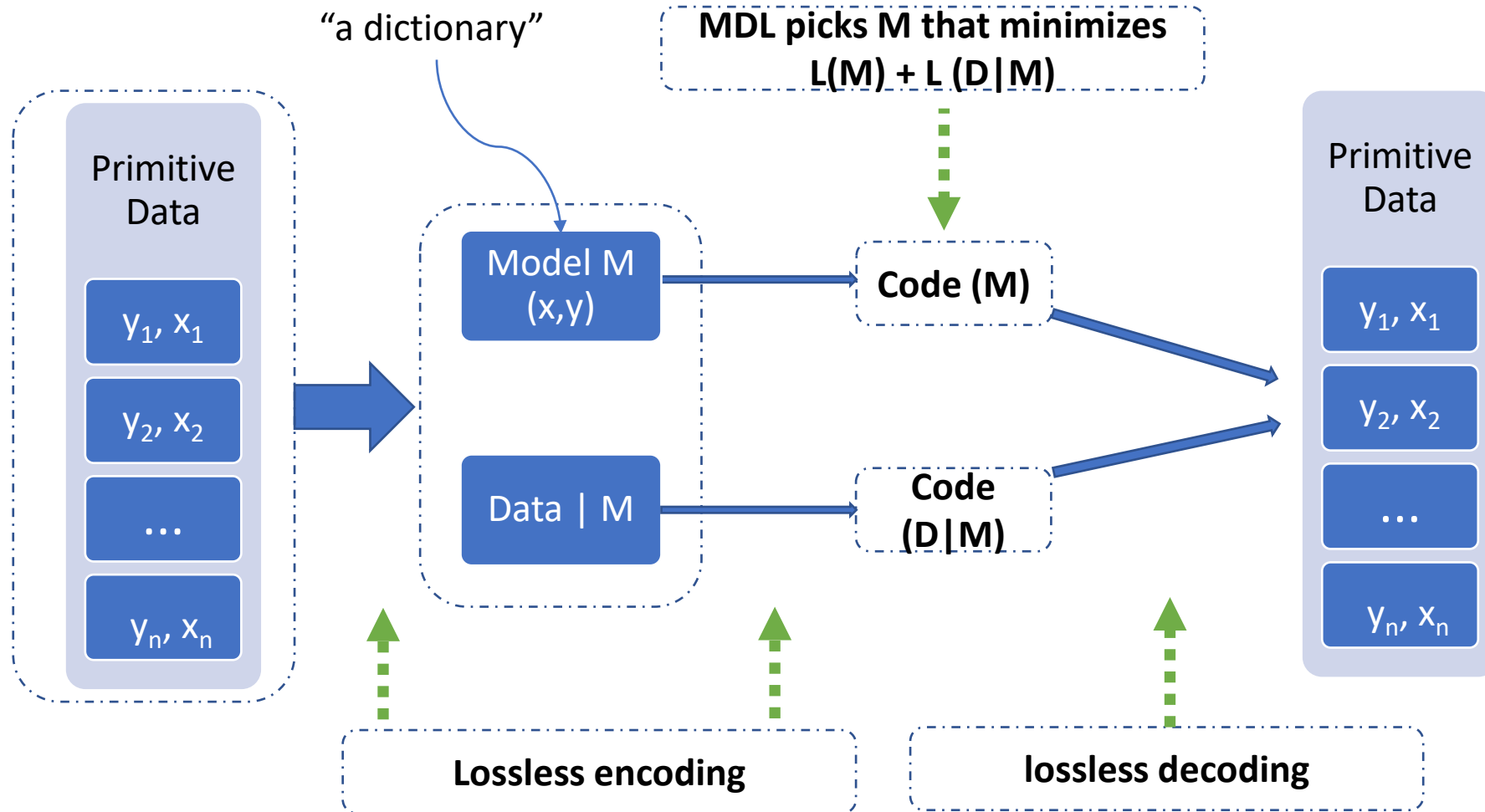
rule 30



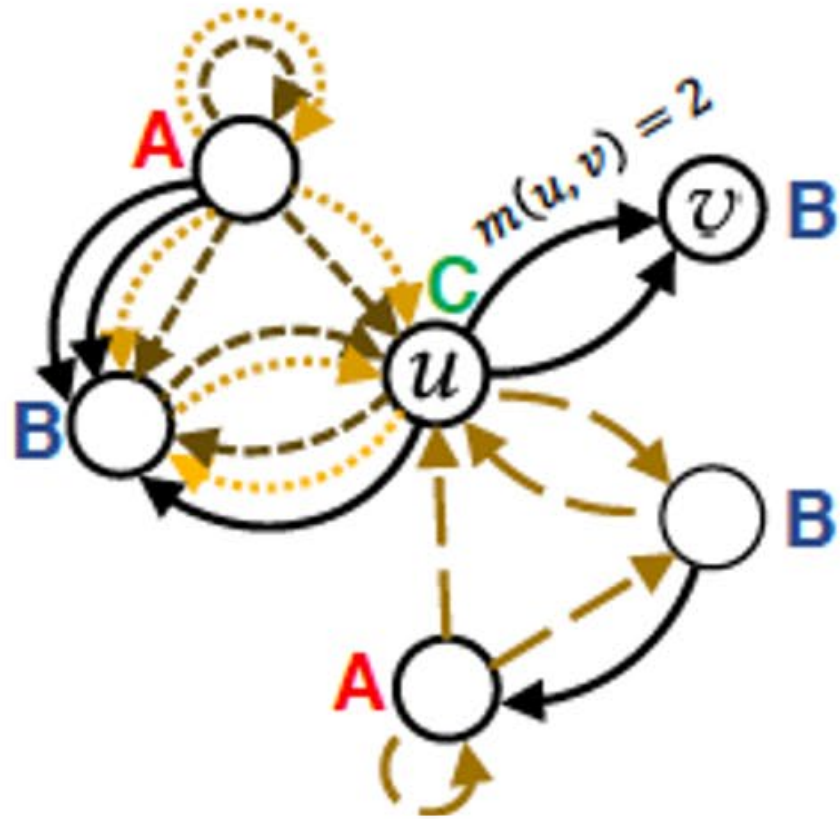
Stephen Wolfram's ECA Rule-30



Minimum Description Length (MDL) Principle: Compress data to recover patterns/models



One transaction graph



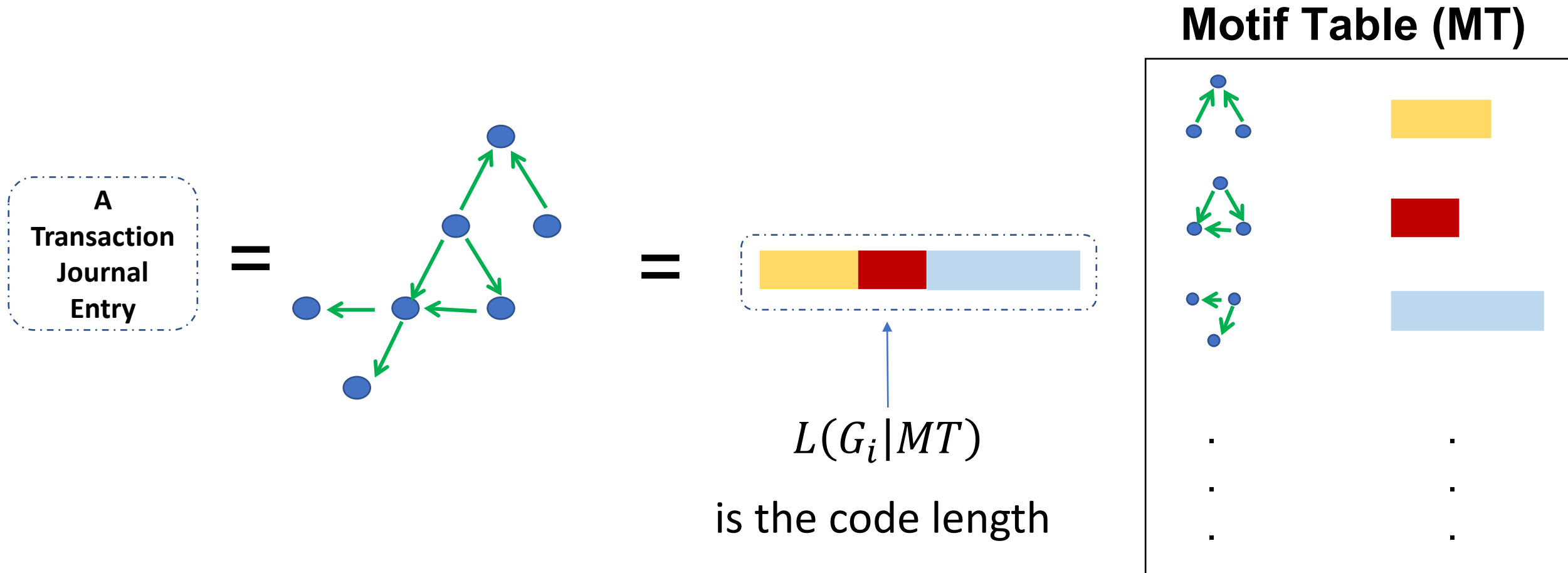
(a)

Motif Table (MT)

g_1		<input type="text"/>	c_1
g_2		<input type="text"/>	c_2
g_3		<input type="text"/>	c_3
g_4		<input type="text"/>	c_4

(b)

Graph Encoding via MDL Principle





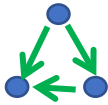



Graph Encoding via MDL Principle

$$\mathcal{G} = G_1 \quad G_2 \quad G_3 \quad \dots \quad G_J$$

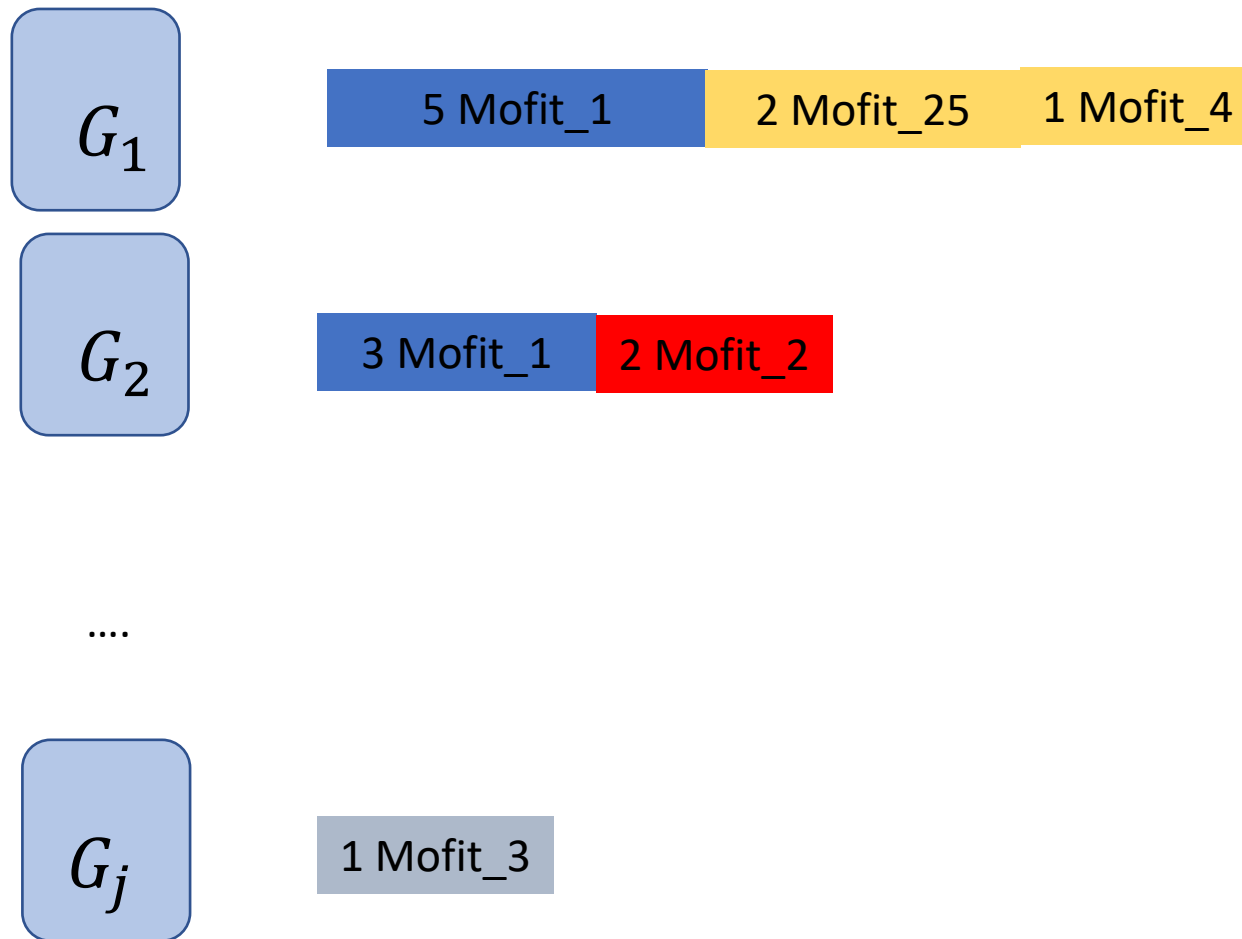
$$\sum_i L(G_i | MT) + L(MT)$$

Minimize by choosing a
best Motif Table

Motif Table (MT)

	
	
	
.	.
.	.
.	.

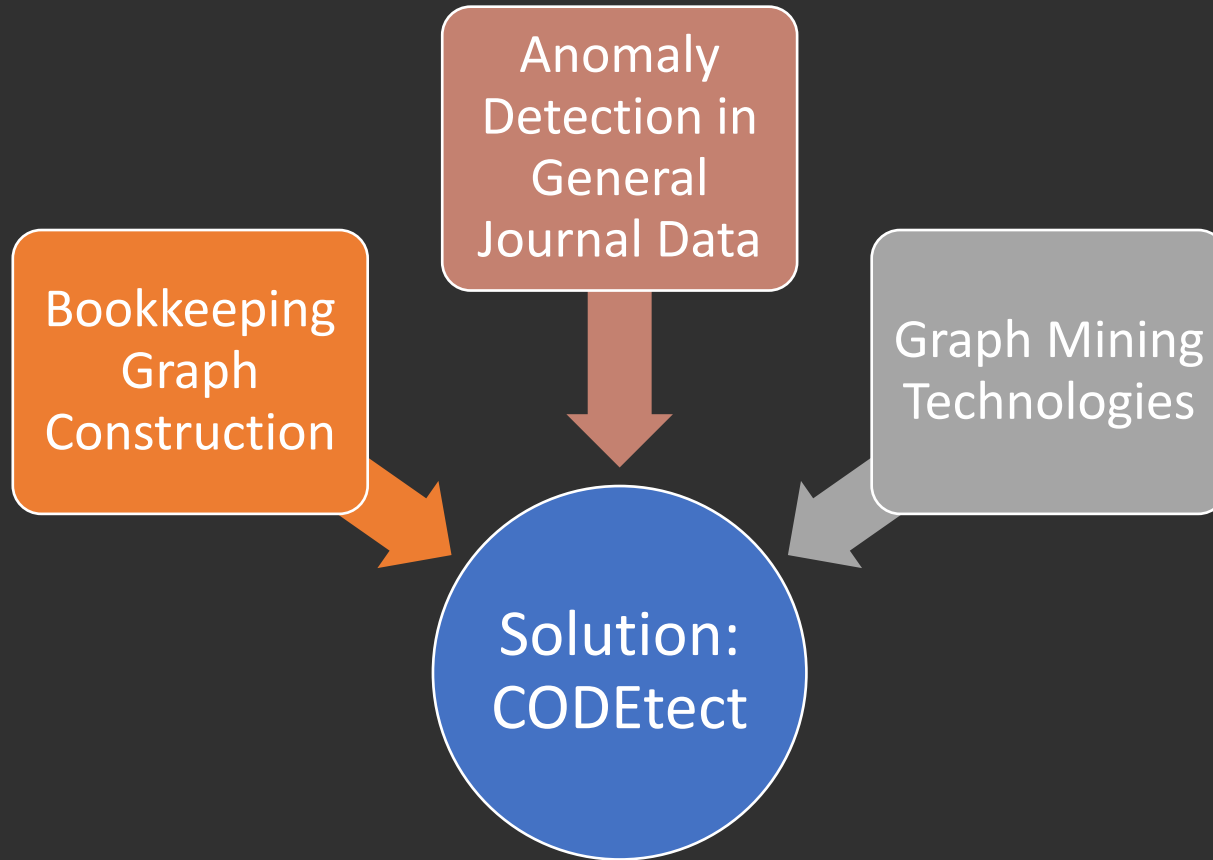
Use Graph Encoding to ID Anomalies



Best Motif Table

.	.
.	.
.	.

Careful Mix Leads to Solution



Broad Lessons

- **Broader View of Research**
 - Get closer to real world problems faced by practitioners and/or policy makers
- **Return to Core Accounting Ideas**
 - Double-entry bookkeeping as account graphs
 - Classification and aggregation
- **Learn from Engineers**
 - Graph theory
 - Graph Mining

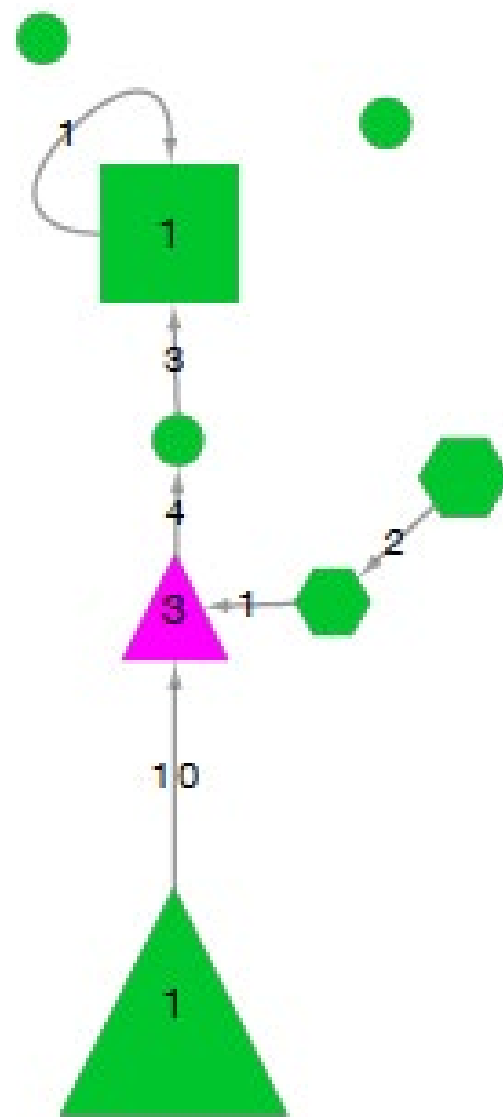
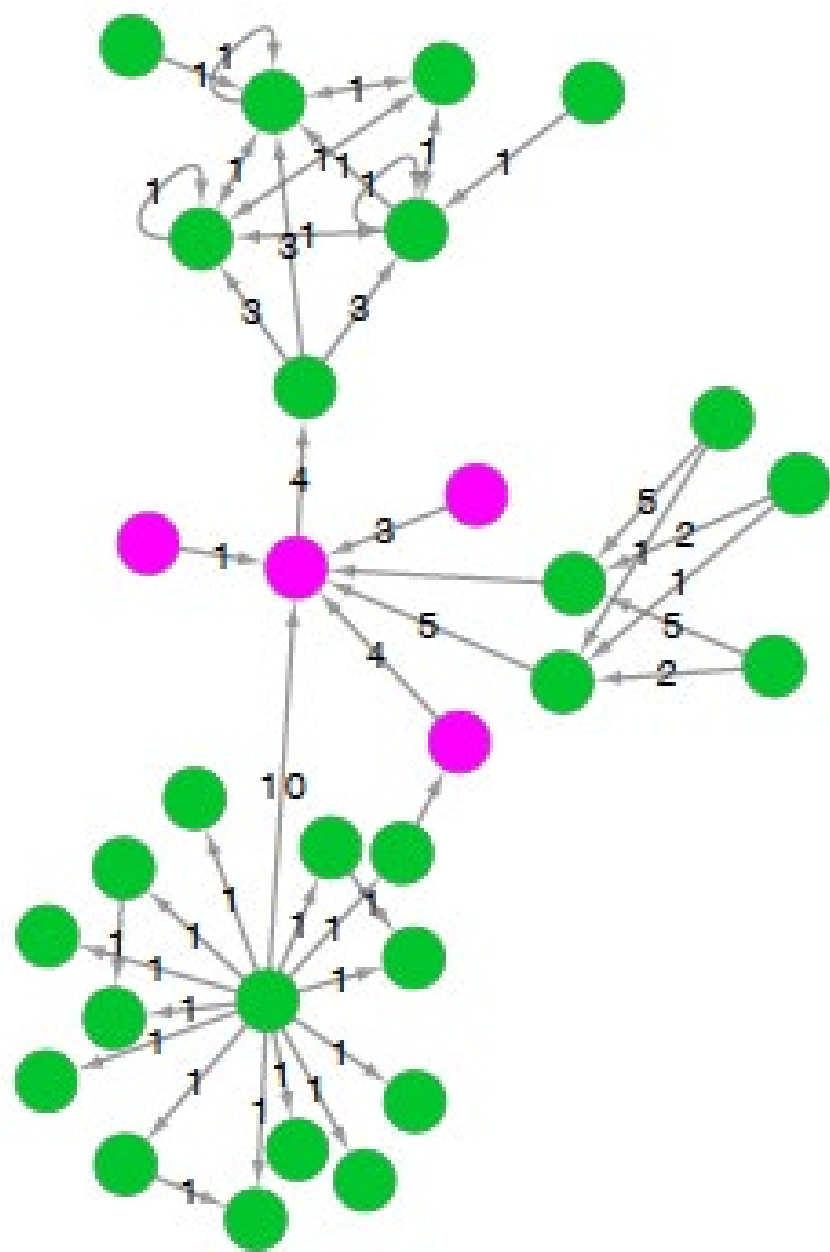
Second USE CASE:

Using Graph Summary to Quantify Aggregation in Bookkeeping Data

Principal Investigators:

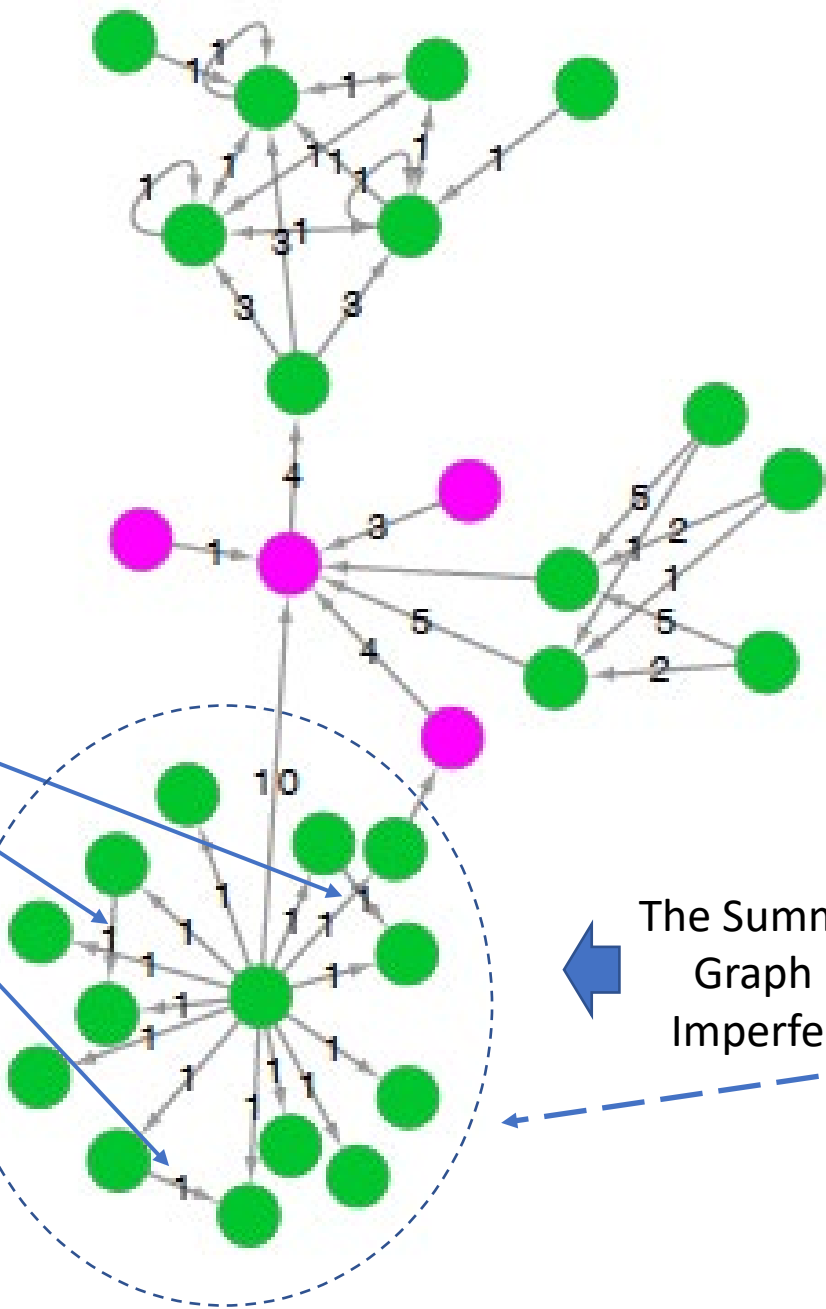
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The Original Graph

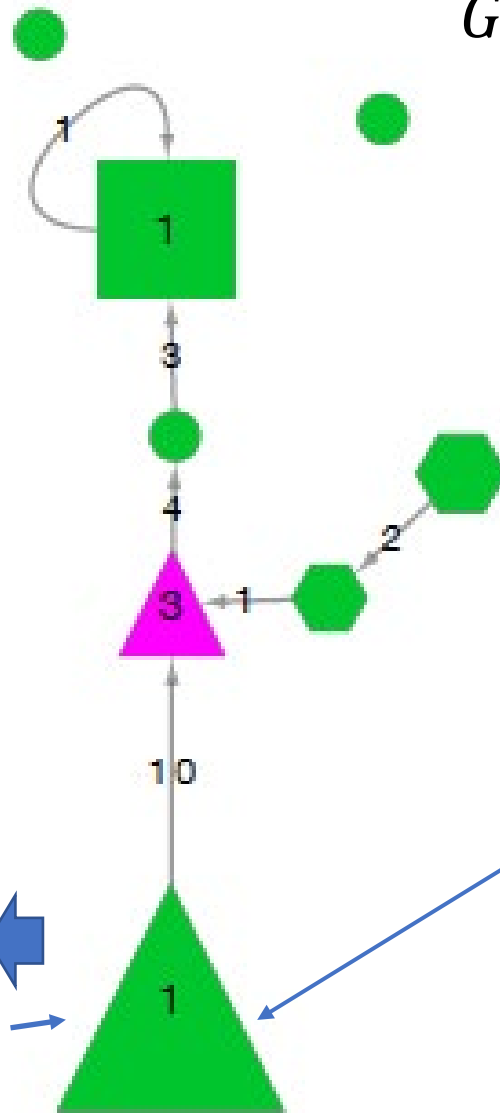
G



Imperfections

The Summary Graph

G_S

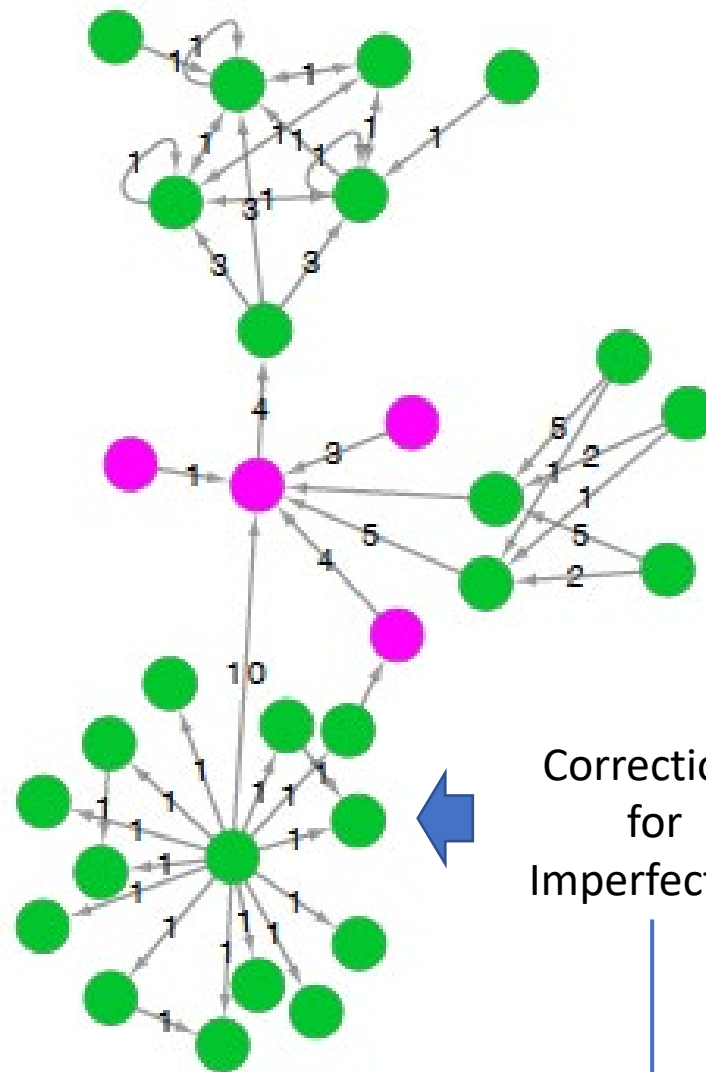


The Summary Graph is Imperfect!

"Triangle" super-node represents a Star

The
Original
Graph

G

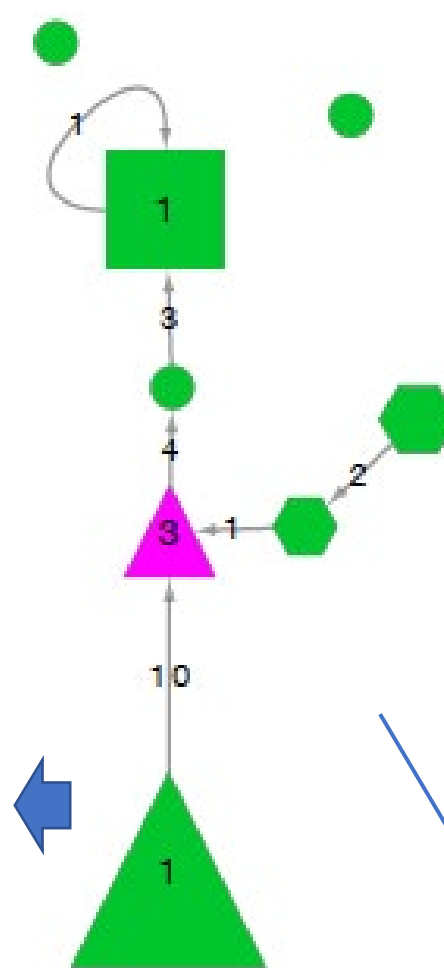


$$L(G) \geq L(G|G_S)$$

Corrections
for
Imperfection

G_S

The
Summary
Graph



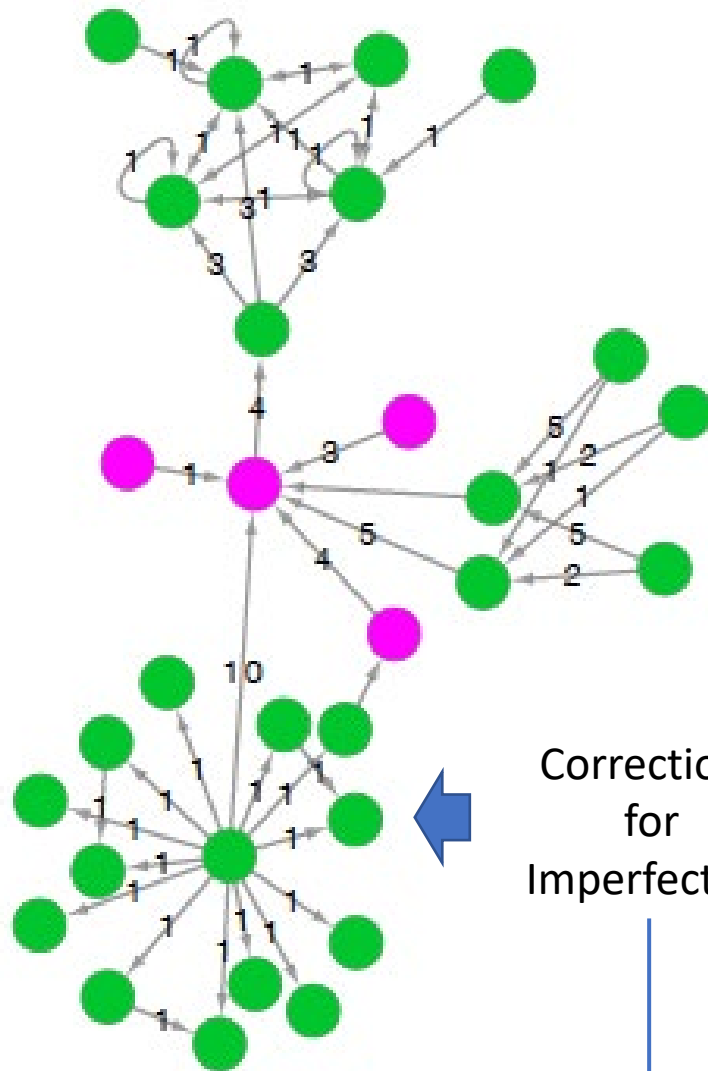
+

Decomposition
without the
corrections

$$L(G_S)$$

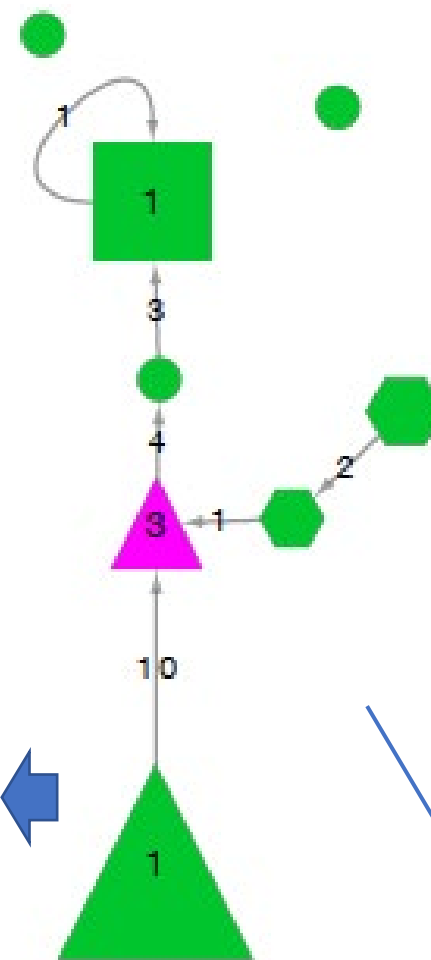
The
Original
Graph

G



G_S

The
Summary
Graph

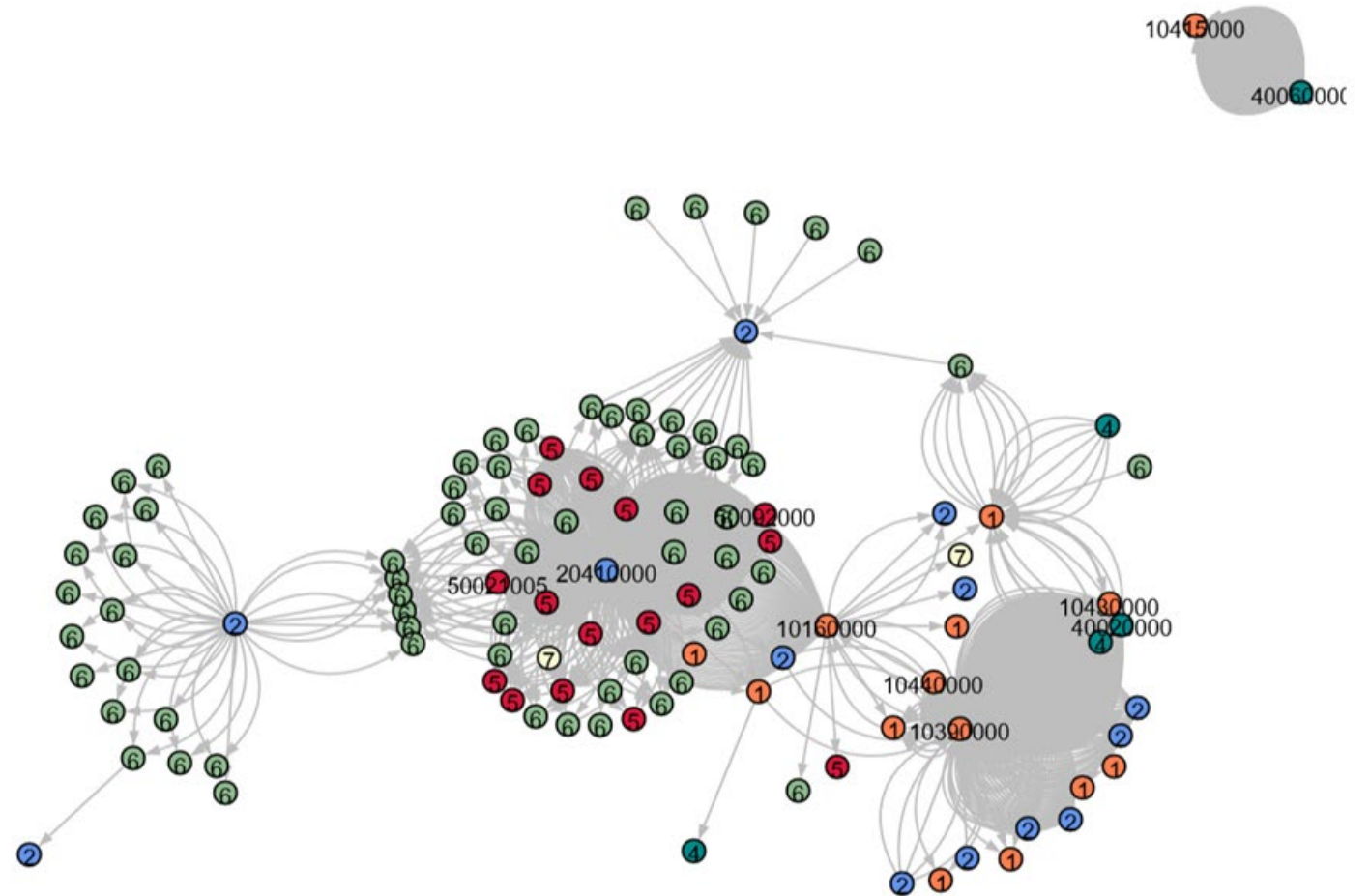


Corrections
for
Imperfection

Decomposition
without the
corrections

$$L(G) \geq L(G|G_S) + L(G_S)$$

Bookkeeping Graph for a 10- day Window



Quantifying Account Classification and Aggregation

We use TG-sum for evaluating a preexisting node labeling, the set of types pre-assigned to the nodes in a graph

Financial Statement Classification (**GAAP**) organizes the individual accounts typical GAAP categories such as inventory, prepaid expense, accounts payable, etc. For each dataset, we use specific classification given the COA given with the data.

Economic Bookkeeping Classification (**EB**) organizes the individual accounts into operating versus financing and long versus short term accounts, all together fourteen categories.

Quantifying Account Classification and Aggregation

Table 4: Evaluating account labelings in financial networks

Company	Labeling		Shuffled	Actual	norm. gain (%)
SH	EB		0.28	0.32	5.6 %
	FS		0.25	0.27	2.7 %
HW	EB		0.36	0.47	17.0 %
	FS		0.16	0.27	13.0 %
KD	EB		0.33	0.42	13.7 %
	FS		0.31	0.39	12.0 %

Many thanks!

- **Hope you learned something more about**
 - **computer science**
 - **graph theory**
 - **accounting**
- **I told you this is an unusual talk**