# Building A Conceptual Framework for

Network Analytics

Minjian Liu (u5506264) Supervised by Dr. Qing Wang

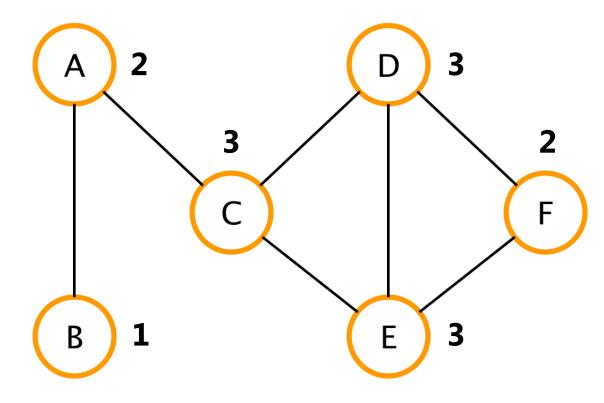


### OUTLINE



#### Ranking: the most important vertex

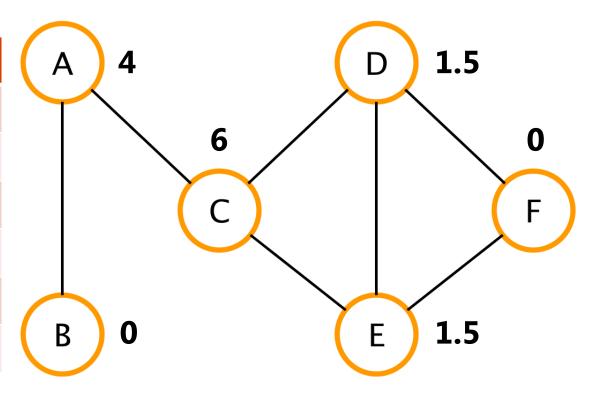
VertexID	Degree
<u>C</u>	<u>3</u>
<u>D</u>	<u>3</u>
<u>E</u>	<u>3</u>
Α	2
F	2
В	1



#### Ranking: the most important vertex

VertexID	Degree	VertexID
<u>C</u>	3	<u>C</u>
D	3	D
E	<u>3</u>	Е
A	2	Α
F	2	F
В	1	В

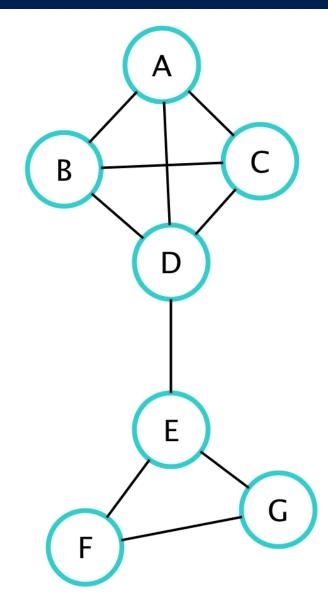
VertexID	Betweenness
<u>C</u>	<u>6</u>
D	1.5
Е	1.5
Α	4
F	0
В	0



### Clustering: a group of vertices

#### Connected Component

ClusterID	Size	Members
1	7	{A, B, C, D, E, F, G}



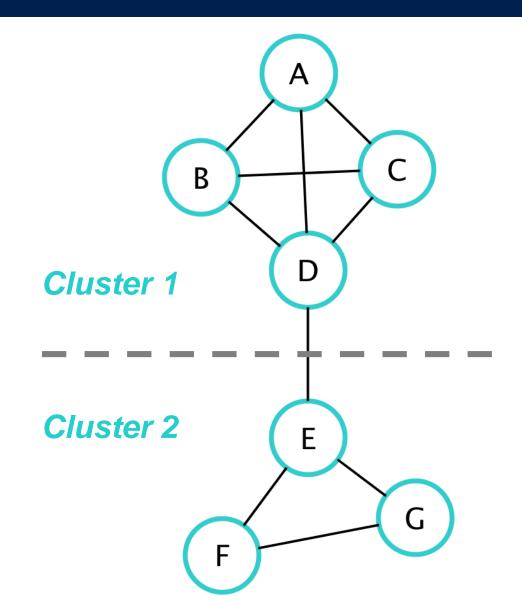
#### Clustering: a group of vertices

#### **Connected Component**

ClusterID	Size	Members
1	7	{A, B, C, D, E, F, G}

#### Community Detection

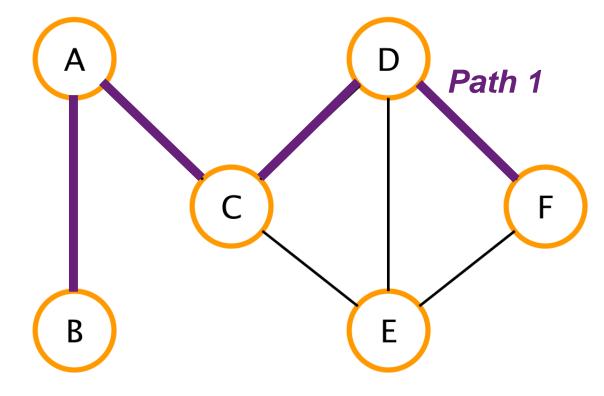
ClusterID	Size	Members
1	4	{A, B, C, D}
2	3	{E, F, G}



### Path Finding: a sequence of vertices

#### A Path between B and F with length 4

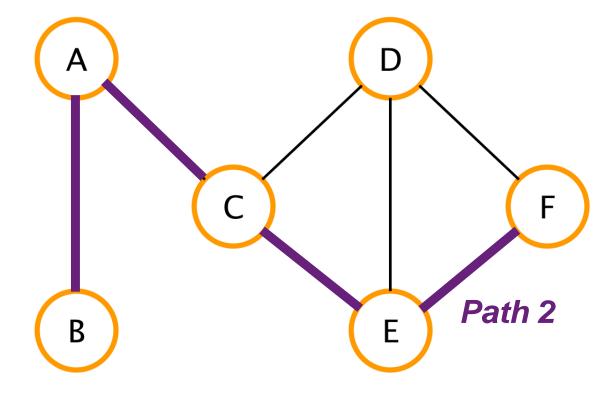
PathID	Length	Path
1	4	<b, a,="" c,="" d,="" f=""></b,>
2	4	<b, a,="" c,="" e,="" f=""></b,>



### Path Finding: a sequence of vertices

#### A Path between B and F with length 4

PathID	Length	Path
1	4	<b, a,="" c,="" d,="" f=""></b,>
2	4	<b, a,="" c,="" e,="" f=""></b,>



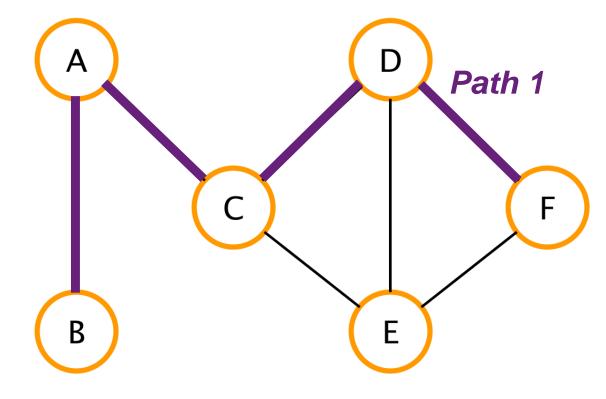
### Path Finding: a sequence of vertices

#### A Path between B and F with length 4

PathID	Length	Path
1	4	<b, a,="" c,="" d,="" f=""></b,>
2	4	<b, a,="" c,="" e,="" f=""></b,>

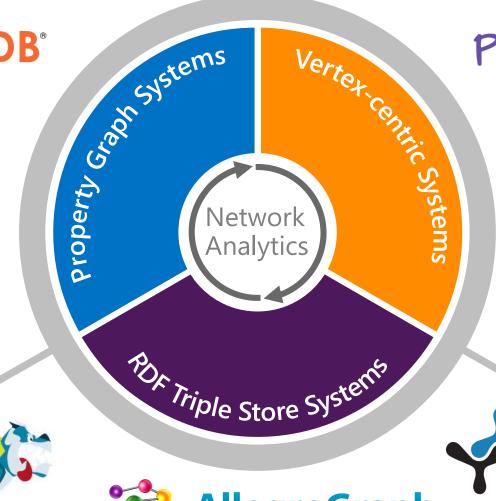
#### A Path between B and F with D in the middle

PathID	Length	Path
1	4	<b, a,="" c,="" d,="" f=""></b,>











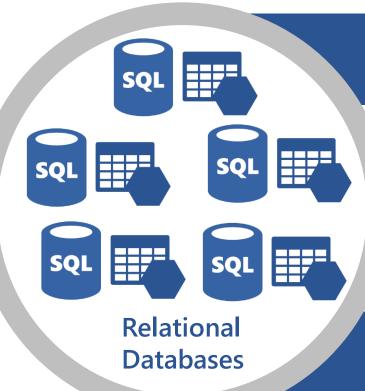




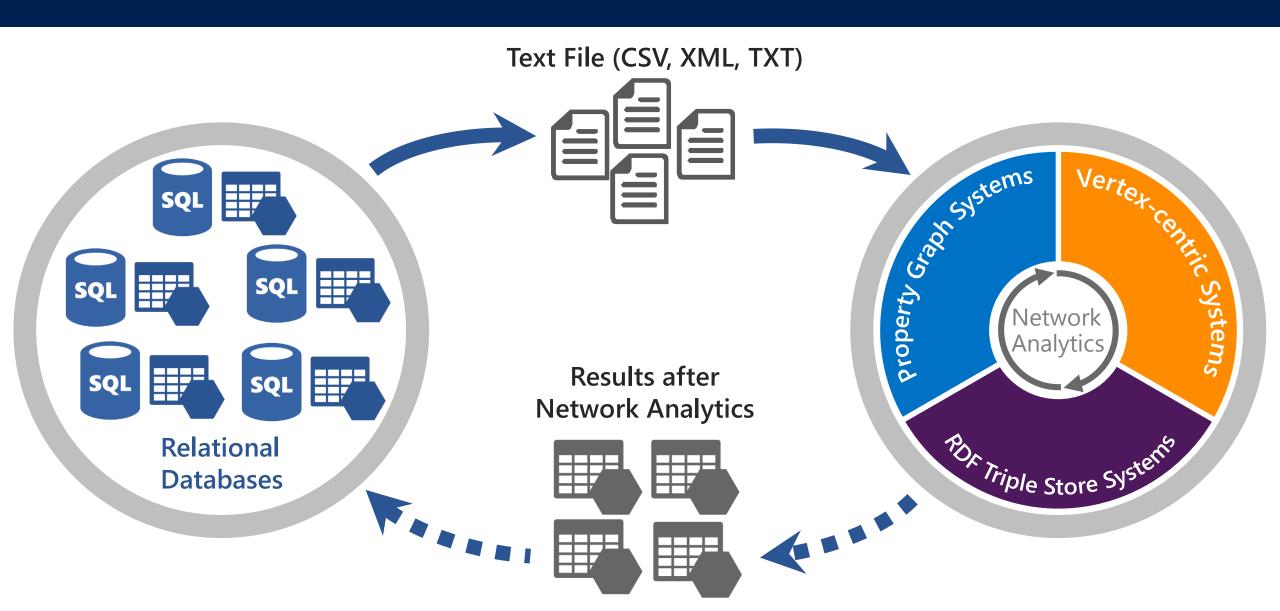








- They are still widely used by enterprises and organizations
  - They are mature and battle-tested with a stable code base
  - People are familiar with using SQL to manipulate data in relational databases
- They provide a number of sophisticated optimisation technologies (e.g. indexing, materialized views)



Data Model

Extends the relational model with network analytics capability

Relational Core



Graphical Views



**RG** Mapper



Query Language



Follows SQL-style syntax to manipulate both schema and network data

Rank

Cluster

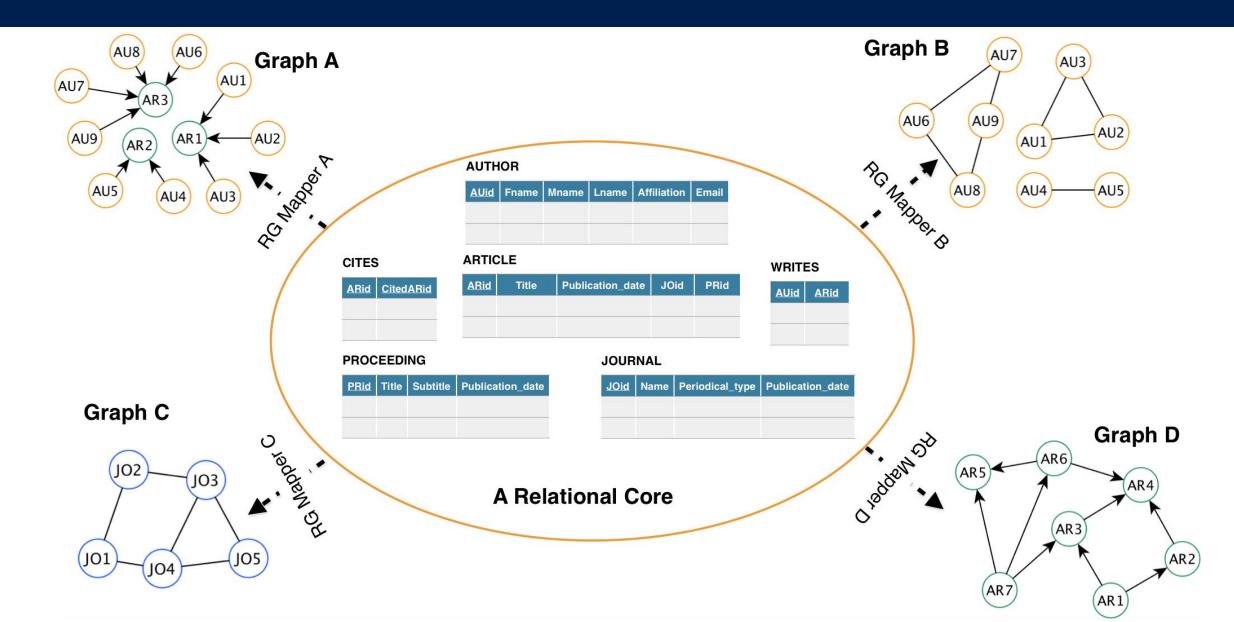
Path







### DATA MODEL





#### Ranking:

Find the most important vertex.



#### **Clustering:**

Find a group of vertices.



#### **Path Finding:**

Find a sequence of vertices.



#### **Create Graph:**

Use a RG mapper to create graph.

Find top 3 influential authors in the co-authorship network.

**SELECT** VertexID, Value

FROM RANK (coauthorship, betweenness)

LIMIT 3;



#### Ranking: Find the most



important vertex.



#### Clustering:

Find a group of vertices.



#### **Path Finding:**

Find a sequence of vertices.



#### **Create Graph:**

Use a RG mapper to create graph.

Find top 3 influential authors in the co-authorship network.

**SELECT** VertexID, Value

**FROM** RANK (coauthorship, betweenness)

LIMIT 3;

Graph Graph Operator Name

VertexID	Value
AU8	6
AU3	1.5
AU9	1.5

degree; indegree outdegree; pagerank betweenness; closeness

Ranking Measure



#### Ranking:

Find the most important vertex.



#### Clustering:

Find a group of vertices.





#### **Path Finding:**

Find a sequence of vertices.



#### **Create Graph:**

Use a RG mapper to create graph.

Find the biggest community that consists of authors who collaborate with each other to publish articles together.

**SELECT** ClusterID, Size, Members

**FROM** CLUSTER (coauthorship, GN)

**ORDER BY Size DESC** 

LIMIT 1;



#### Ranking: Find the most

important vertex.



#### Clustering:

Find a group of vertices.





#### **Path Finding:**

Find a sequence of vertices.



#### **Create Graph:**

Use a RG mapper to create graph.

Find the biggest community that consists of authors who collaborate with each other to publish articles together.

**SELECT** ClusterID, Size, Members

FROM CLUSTER (coauthorship, GN)

**ORDER BY Size DESC** 

LIMIT 1;

**Graph Name** 

Graph Operator

CC (Connected Component)
SCC (Strongly Connected Component)
GN (Girvan-Newman algorithm)
CNM (Clauset-Newman-Moore Algorithm)
MC (Monte Carlo Algorithm)

ClusterID	Size	Members
2	4	{AU1, AU3, AU8, AU9}

Clustering Algorithm



### Ranking: Find the most

important vertex.



#### Clustering:

Find a group of vertices.



#### **Path Finding:**

Find a sequence of vertices.



#### **Create Graph:**

Use a RG mapper to create graph.

Find shortest paths between two authors V1 and V2, author V1 is affiliated at ANU and author V2 is affiliated at NICTA.

```
SELECT PathID, Length, Path
FROM PATH (coauthorship, V1// V2)
WHERE V1 AS

(
SELECT AUID FROM AUTHOR
WHERE Affiliation = 'ANU'
) AND V2 AS

(
SELECT AUID FROM AUTHOR
WHERE Affiliation = 'NICTA'
)
ORDER BY Length ASC;
```



#### Ranking: Find the most

important vertex.



#### **Clustering:**

Find a group of vertices.



#### Path Finding:

Find a sequence

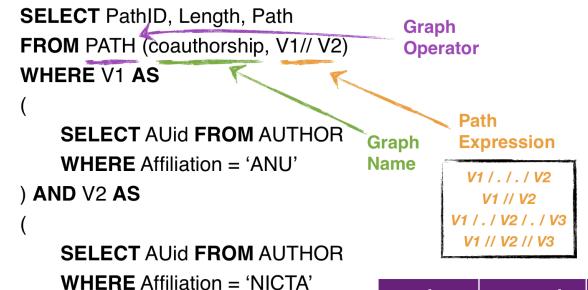


of vertices.

#### **Create Graph:**

Use a RG mapper to create graph.

Find shortest paths between two authors V1 and V2, author V1 is affiliated at ANU and author V2 is affiliated at NICTA.



**ORDER BY** Length **ASC**;

PathID	Length	Path		
1	1	<au1, au8=""></au1,>		



#### Ranking:

Find the most important vertex.



#### Clustering:

Find a group of vertices.



#### **Path Finding:**

Find a sequence of vertices.



#### **Create Graph:**

Use a RG mapper to create graph.



#### Create a materialized graph for co-authorship network.

```
CREATE UNGRAPH coauthorship AS

(

SELECT w1.AUid, w2.AUid
FROM WRITES as w1, WRITES as w2
WHERE w1.ARid = w2.ARid AND w1.AUid != w2.AUid
);
```



#### Ranking:

Find the most important vertex.



#### Clustering:

Find a group of vertices.



#### **Path Finding:**

Find a sequence of vertices.



#### **Create Graph:**

Use a RG mapper to create graph.



Create a citation graph on-the-fly for ranking.

**SELECT** VertexID, Value

**FROM** RANK (citation, indegree)

WHERE citation IS DIGRAPH AS

SELECT ARid, CitedARid FROM CITES

UNGRAPH (Undirected Graph)
DIGRAPH (Directed Graph)

**Graph Type** 

**Graph Name** 



### QUERY ENGINE

**Query Console Query Analyser & Rewriter Relational Query** Rank Query Parser Executor Plan Cost Cluster **Estimator** Generator Query Executor **Relational Query Optimizer** Path **Relational Query** Query **Executor Executor** 

**Relation-Graph Engine (RG Engine)** 

#### **Query Console**

- Allows user enter queries
- Shows the query results

#### **Query Analyser & Rewriter**

- Checks whether or not the query syntax is correct
- Differentiates graph sub-queries and relational sub-queries
- Rewrites graph sub-queries

#### **Graph Query Components**

- Three query executors execute three types of network analytics tasks (ranking, clustering, path finding).
- They transform the results into temporary tables stored in data storage.

### QUERY ENGINE

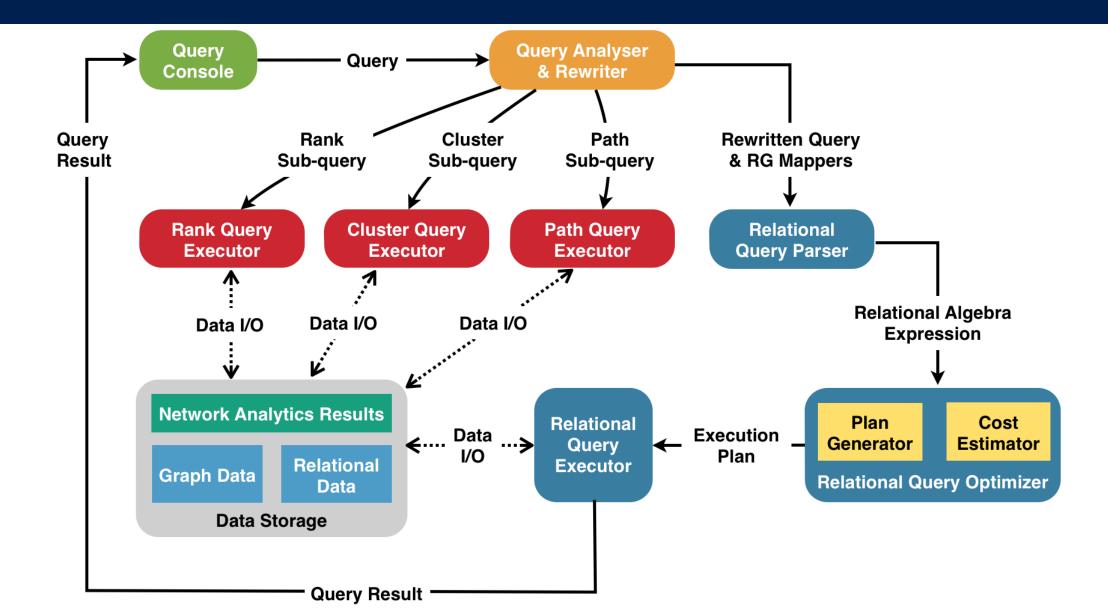
**Query Console Query Analyser & Rewriter** Rank **Relational Query** Query Parser **Executor** Plan Cost Cluster **Estimator** Generator Query Executor **Relational Query Optimizer** Path **Relational Query** Query **Executor Executor** 

#### **Relational Query Components**

- Parser checks if operators are consistent with data types and converts the query into relational algebra expression.
- Optimiser generates alternative execution plans based on the relational algebra expression, estimates the cost for a subset of plans and chooses the best plan.
- Executor executes the best plan and returns results to the query console

**Relation-Graph Engine (RG Engine)** 

### QUERY ENGINE



 We adopt 3 graph analysis tools including SNAP, NetworkX, Graph-tool as algorithm support for our query executors.

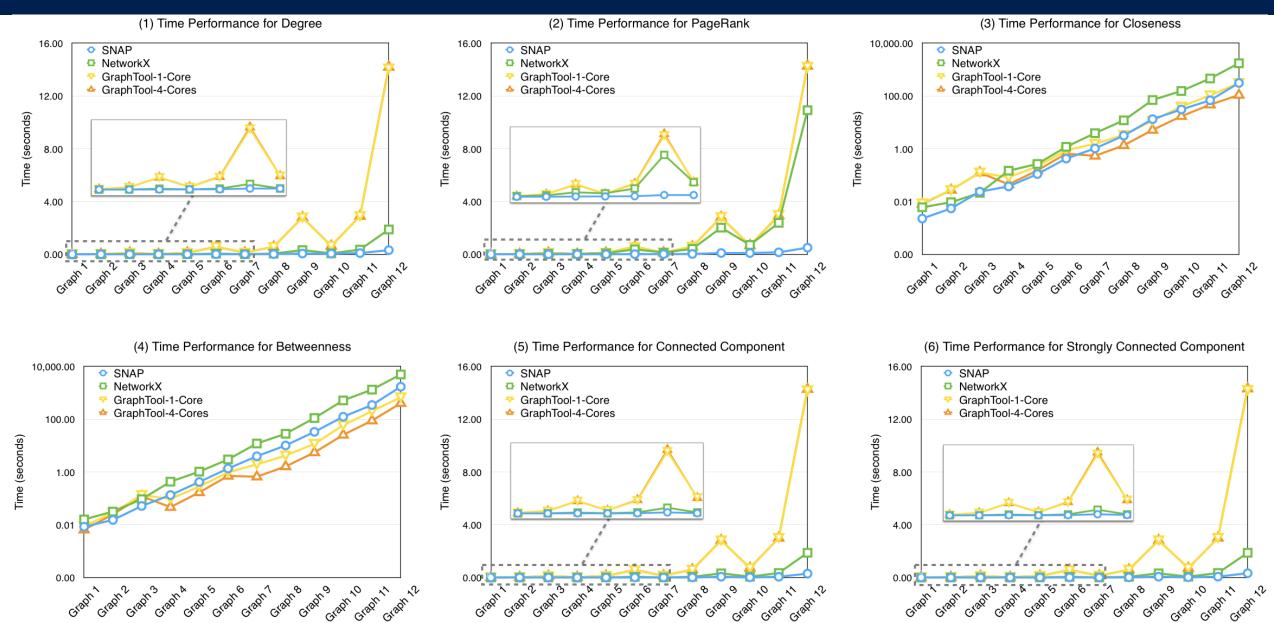
#### **Functions of Graph Analysis Tools**

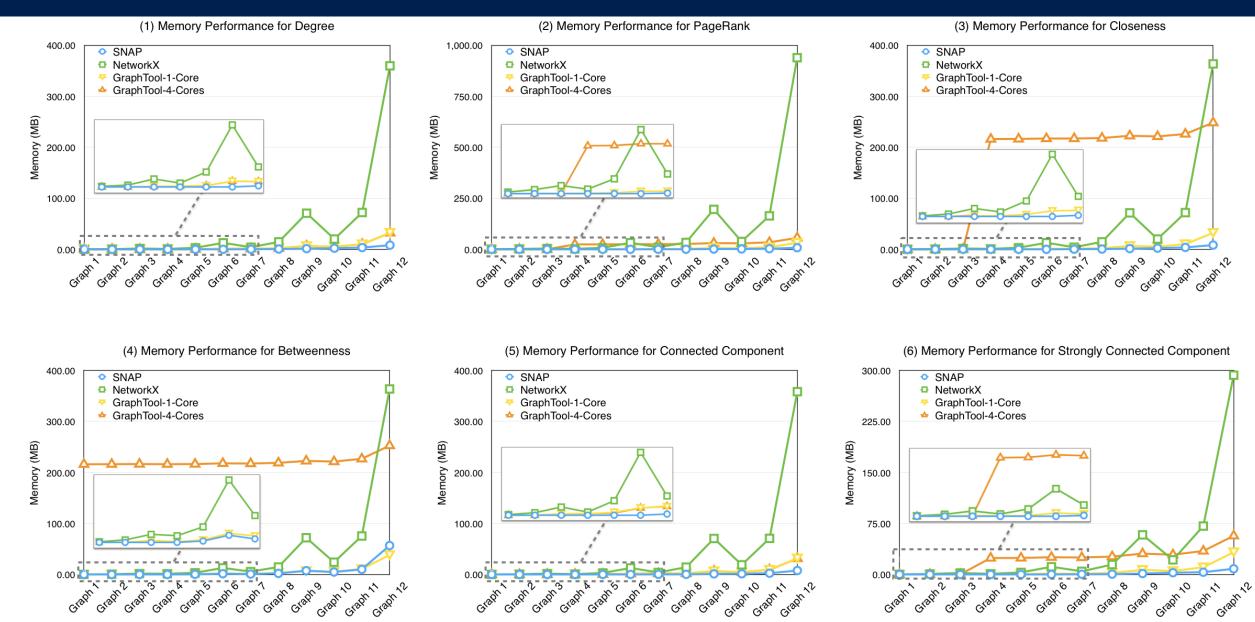
	Algorithms	SNAP	NetworkX	GraphTool
Ranking	Degree	✓	✓	✓
	Pagerank	✓	✓	✓
	Betweenness	✓	✓	✓
	Closeness	✓	✓	✓
Clustering	Connected Component(CC)	✓	✓	✓
	Strongly Connected Component(SCC)	✓	✓	✓
	Girvan-Newman (GN)	✓	_	_
	Clauset-Newman-Moore (CNM)	✓	_	_
	Monte Carlo (MC)	_	_	✓
Path Finding	Shortest Path	_	✓	_
	Path with Specific Length	_	✓	_

- We adopt 3 graph analysis tools including SNAP, NetworkX, Graph-tool as algorithm support for our query executors.
- For the first part of experiment, we use Erdos-Renyi method to create random graph as the experiment input.

#### **Erdos-Renyi Random Graph**

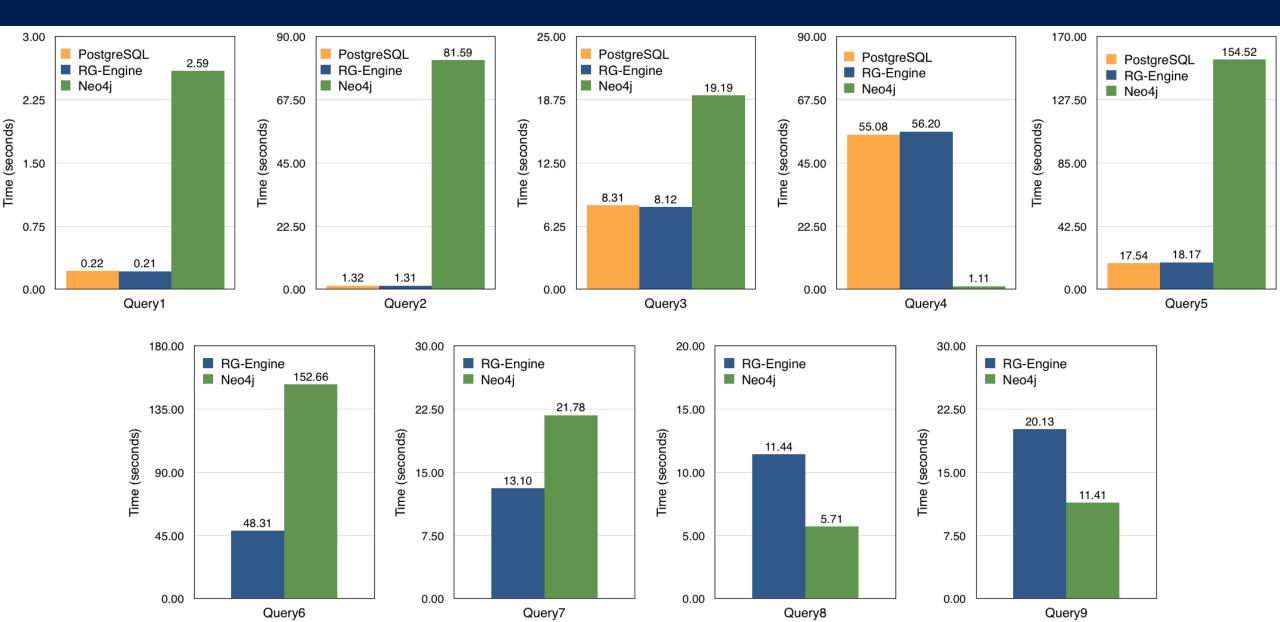
	Number of Vertices	Number of Edges	Size(KB)
Graph 1	100	200	1
Graph 2	100	1,000	6
Graph 3	100	5,000	29
Graph 4	500	1,000	8
Graph 5	500	5,000	38
Graph 6	500	25,000	189
Graph 7	2,500	5,000	46
Graph 8	2,500	25,000	228
Graph 9	2,500	125,000	1,100
Graph 10	12,500	25,000	256
Graph 11	12,500	125,000	1,300
Graph 12	12,500	625,000	6,400





- Queries 1 3 are traditional relational queries including join operations and aggregate operations.
- Queries 4 10 are about some typical network analytics tasks including pattern matching, triangle counting, pagerank centrality, connected component, path finding and community detection
- Queries 11 12 are advanced queries that are nested with two different types of network analytics tasks.

		PSQL	RG-Engine	Neo4j
Query 1	Aggregate Operation + Set Operation	✓	✓	✓
Query 2	Aggregate Operation + Sorting Operation	<b>√</b>	✓	✓
Query 3	Aggregate Operation  + Sorting Operation  + Join Operation	✓	✓	✓
Query 4	Pattern Mathcing	✓	✓	✓
Query 5	Triangle Counting	✓	✓	✓
Query 6	PageRank Centrality	_	✓	✓
Query 7	Connected Component	_	✓	✓
Query 8	Path Finding	_	✓	✓
Query 9	Shortest Path	_	✓	✓
Query 10	Community Detection	_	✓	_
Query 11	PageRank Centrality + Connected Component	_	✓	-
Query 12	PageRank Centrality + Path Finding	_	✓	_



# Thank You! Q&A



## Contact: Minjian Liu



0426839321



378610682



u5506264@anu.edu.au

Some of icons are downloaded from www.rapidbbs.cn through its membership services.

The source code and query samples are available at https://gitlab.com/Minken/COMP8800\_Computing\_Research\_Project.git