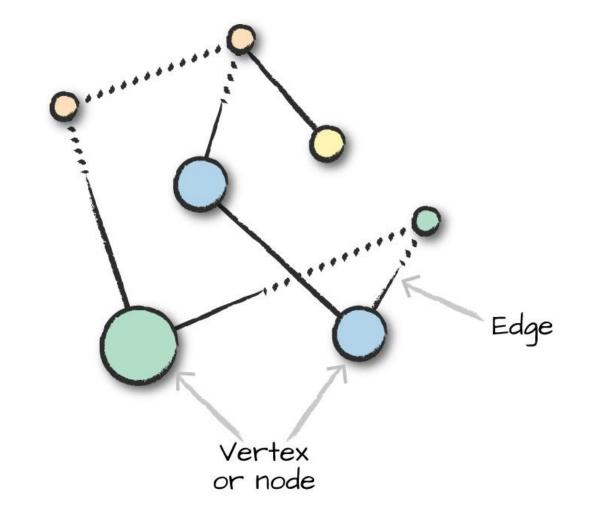
### **Graph Analytics in Spark**

### **Graph analytics: Introduction**

### **Graph analytics**

- Graphs are data structures composed of nodes and edges
  - Nodes/vertexes are denoted as V={v<sub>1</sub>, v<sub>2</sub>, ..., v<sub>n</sub>} and edges are denoted as E={e<sub>1</sub>, e<sub>2</sub>, ..., e<sub>n</sub>}
  - Graph analytics is the process of analyzing relationships between vertexes and edges

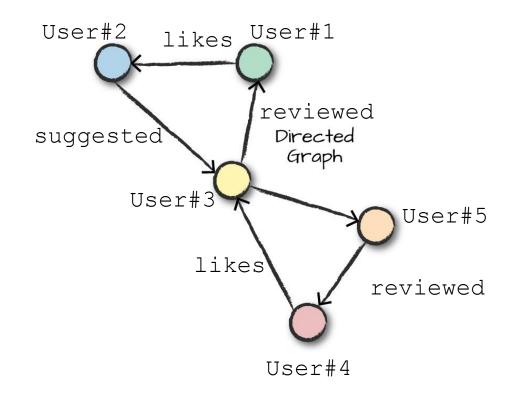
### **Graph analytics**



### Vertexes, edges and weights

- Graphs are undirected if edges do not have a direction
- Otherwise they are called directed graphs
- Vertexes and edges can have data associated with them
  - weight/label
    - e.g., an edge weight may represent the strength of the relationship
    - e.g., a vertex label may be the string associated with the name of the vertex

### Vertexes, edges and weights

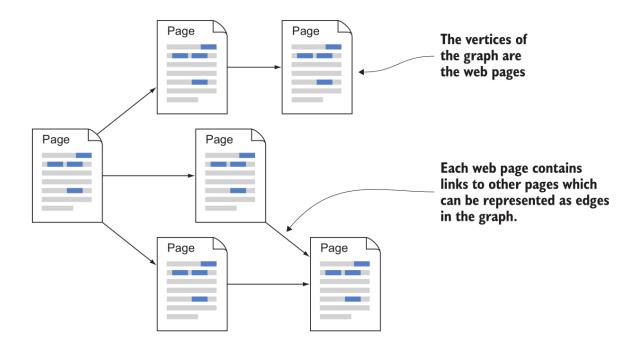


# Why graph analytics?

- Graphs are natural way of describing relationships
- Practical example of analytics over graphs
  - Ranking web pages (Google PageRank)
  - Detecting group of friends
  - Determine importance of infrastructure in electrical networks

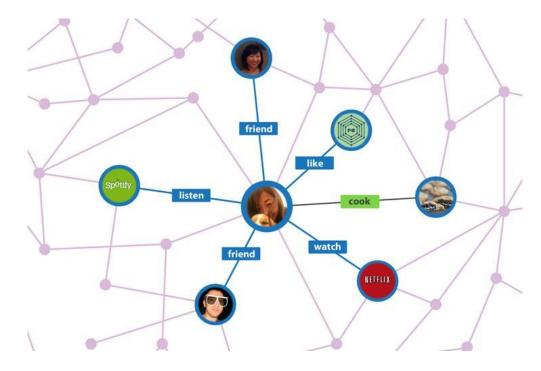
### Graph structure in the web

Importance and rank of web pages



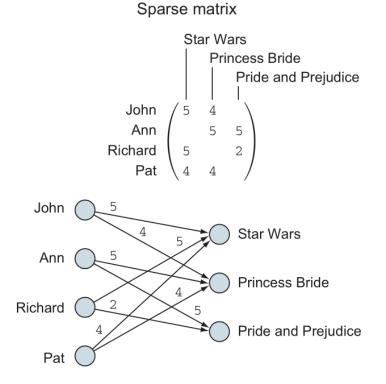
### Graph structure in the web

Social network structure and web usage



### Graph structure in the web

### Movies watched by users

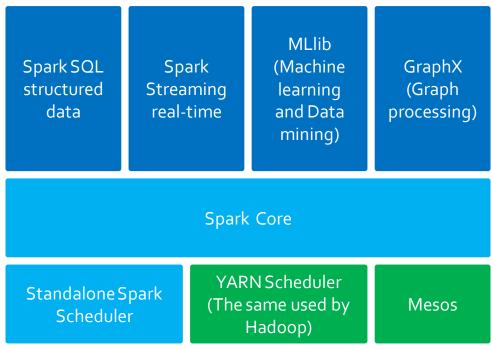


# Spark GraphX and GraphFrames



### GraphX

- Spark RDD-based library for performing graph processing
   Core part of Spark
- Core part of Spark



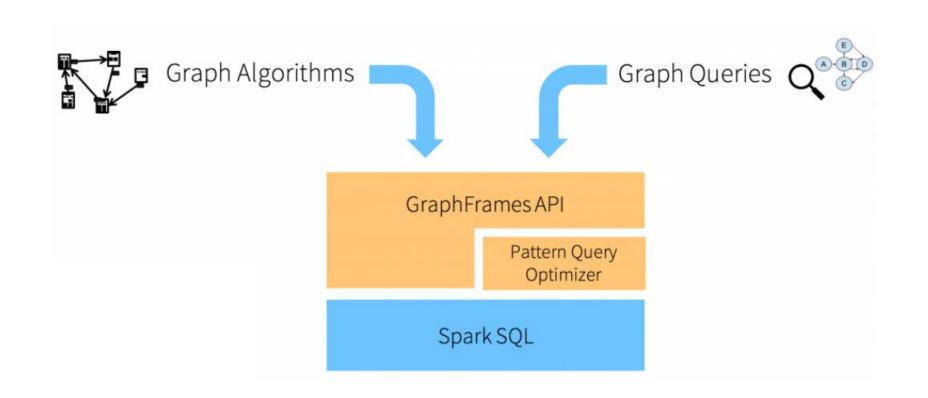
### GraphX

- Low level interface with RDD
- Very powerful
  - Many application and libraries built on top of it
- However, not easy to use or optimize
- No Python version of the APIs

### GraphFrames

- Library DataFrame-based for performing graph processing
- Spark external package built on top of GraphX
  - https://graphframes.github.io/graphframes/docs/\_site/index.html

### GraphFrames



# Building and querying graphs with GraphFrames

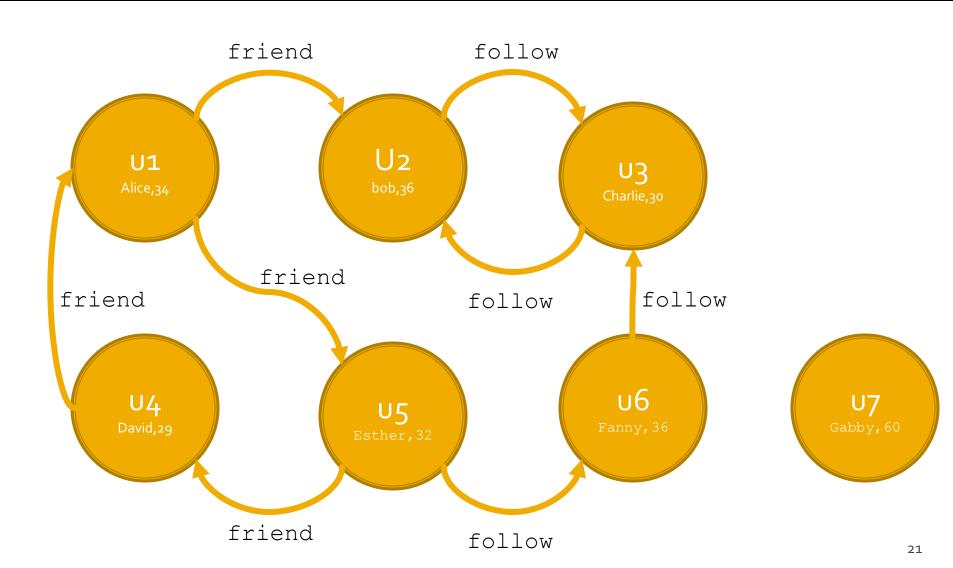
- Define vertexes and edges of the graph
  - Vertexes and edges are represented by means of records inside DataFrames with specifically named columns
    - One DataFrame for the definition of the vertexes of the graph
    - One DataFrame for the definition of the edges of the graph

- The DataFrames that are used to represent nodes/vertexes
  - Contain one record per vertex
  - Must contain a column named "id" that stores unique vertex IDs
  - Can contain other columns that are used to characterize vertexes

- The DataFrames that are used to represent edges
  - Contain one record per edge
  - Must contain two columns "src" and "dst" storing source vertex IDs and destination vertex IDs of edges
  - Can contain other columns that are used to characterize edges

- Create a graph of type graphframes.graphframe.GraphFrame by invoking the constructor GraphFrame(v,e)
  - V
    - The DataFrame containing the definition of the vertexes
  - e

The DataFrame containing the definition of the edges
 Graphs in graphframes are directed graphs



#### Vertex DataFrame

+-	+	+-	+
	id	name a	age
+-	+	+-	+
	u1	Alice	34
	u2	Bob	36
	u3	Charlie	30
	u4	David	29
	u5	Esther	32
	u6	Fanny	36
	u7	Gabby	60
+-	+	+-	+

#### Edge DataFrame

++	+	+			
src	src dst relationship				
++	+	+			
u1	u2	friend			
u2	u3	follow			
u3	u2	follow			
u6	u3	follow			
u5	u6	follow			
u5	u4	friend			
u4	u1	friend			
u1	u5	friend			
++	+	+			

from graphframes import GraphFrame

# Create the graph
g = GraphFrame(v, e)

### Directed vs undirected edges

- In undirected graphs the edges indicate a twoway relationship (each edge can be traversed in both directions)
- In GraphX you could use to\_undirected() to create an undirected copy of the Graph
- Unfortunately GraphFrames does not support it
  - You can convert your graph by applying a flatMap function over the edges of the directed graph that creates symmetric edges and then create a new GraphFrame

### Cache graphs

- As with RDD and DataFrame, you can cache graphs in GraphFrame
  - Convenient if the same (complex) graph result of (multiple) transformations is used multiple times in the same application
  - Simply invoke cache() on the GraphFrame you want to cache
    - It persists the DataFrame-based representation of vertexes and edges of the graph

### Querying the graph

- Some specific methods are provided to execute queries on graphs
  - filterVertices(condition)
  - filterEdges(condition)
  - dropIsolatedVertices()
- The returned result is the filtered version of the input graph

### **Querying the graph: filterVertices**

### filterVertices(condition)

- condition contains an SQL-like condition on the values of the attributes of the vertexes
  - E.g., "age>35"
- Selects only the vertexes for which the specified condition is satisfied and returns a new graph with only the subset of selected vertexes

# **Querying the graph: filterEdges**

### filterEdges(condition)

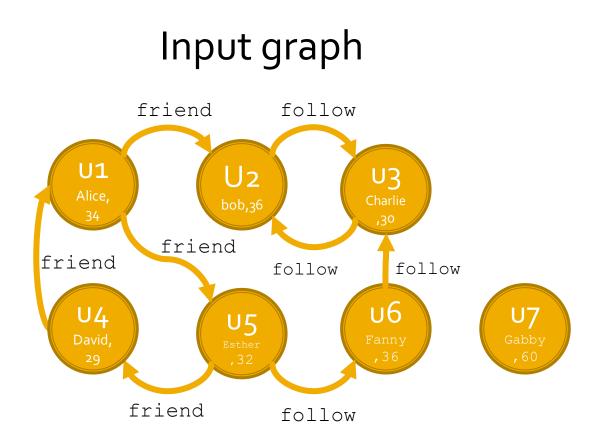
- condition contains an SQL-like condition on the values of the attributes of the edges
  - E.g., "relationship='friend' "
- Selects only the edges for which the specified condition is satisfied and returns a new graph with only the subset of selected edges

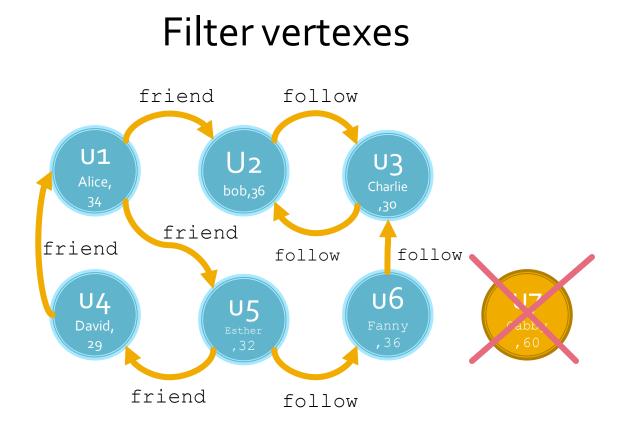
### Querying the graph: dropIsolatedVertices

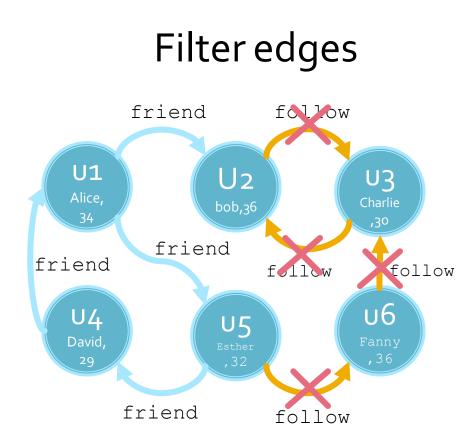
### dropIsolatedVertices()

 Drops the vertexes that are not connected with any other node and returns a new graph without the dropped nodes

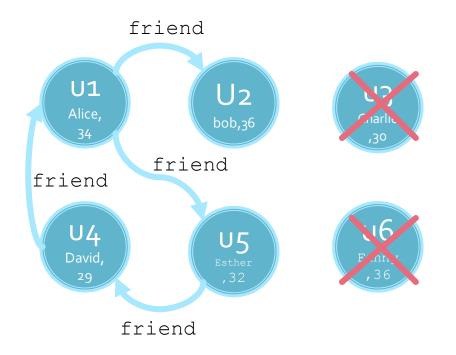
- Given the input graph, create a new subgraph including
  - Only the vertexes associated with users characterized by age between 29 and 50
  - Only the edges representing the friend relationship
  - Drop isolated vertexes



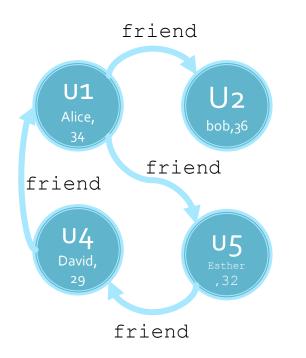


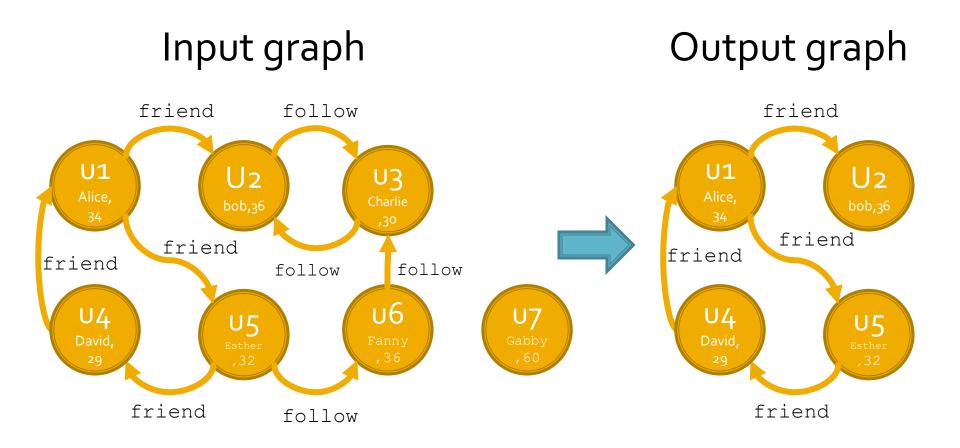


### Drop isolated vertexes



### Output graph





from graphframes import GraphFrame

# Create the graph
g = GraphFrame(v, e)

selectedUsersandFriendRelGraph = g\
.filterVertices("age>=29AND age<=50")\
.filterEdges("relationship='friend'")
. dropIsolatedVertices()</pre>

# **Querying the graph**

- Given a GraphFrame, we can easily access its vertexes and edges
  - g.vertices returns the DataFrame associated with the vertexes of the input graph
  - g.edges returns the DataFrame associated with the edges of the input graph

## **Querying the graph**

- All the standard DataFrame transformations/ actions are available also for the DataFrames that are used to store vertexes and edges
  - For example, the number of vertexes and the number of edges can be computed by invoking the count() action on the DataFrames vertices and edges, respectively

- Given the input graph
  - Count how many vertexes and edges has the graph
  - Find the smallest value of age (i.e., the age of the youngest user in the graph)
  - Count the number of edges of type "follow" in the graph

from graphframes import GraphFrame

# Create the graph
g = GraphFrame(v, e)

# Count how many vertexes and edges has the graph print("Number of vertexes: ",g.vertices.count()) print("Number of edges: ",g.edges.count())

# Print on the standard output the smallest value of age # (i.e., the age of the youngest user in the graph) g.vertices.agg({"age":"min"}).show()

# Print on the standard output
# the number of "follow" edges in the graph.
numFollows = g.edges.filter("relationship = 'follow' ").count()

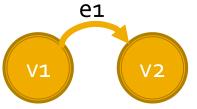
print(numFollows)

# **Motif finding**

- Motif finding refers to searching for structural patterns in graphs
- A simple Domain-Specific Language (DSL) is used to specify the structure of the patterns we are interested in
  - The paths/subgraphs in the graph matching the specified structural pattern are selected

- The basic unit of a pattern is a connection between vertexes
  - (v1) [e1] -> (v2)

means

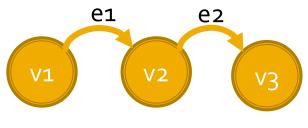


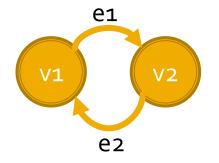
- An arbitrary edge [e1] from an arbitrary vertex (v1) to another arbitrary vertex (v2)
- Edges are denoted by square brackets
  - [e1]
- Vertexes are expressed by round brackets
  - (V1), (V2)

- Patterns are chains of basic units
  - (v1) [e1] -> (v2); (v2) [e2] -> (v3)

means

- An arbitrary edge from an arbitrary vertex v1 to another arbitrary vertex v2 and another arbitrary edge from v2 to another arbitrary vertex v3
  - v3 and v1 can be the same vertex

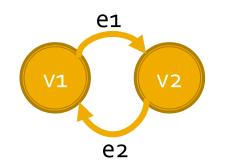




- The same vertex name is used in a pattern to have a reference to the same vertex
  - (v1) [e1] -> (v2); (v2) [e2] -> (v1)

means

 An arbitrary edge from an arbitrary vertex v1 to another arbitrary vertex v2 and vice-versa



- It is acceptable to omit names for vertices or edges in patterns when not needed
  - (v1)-[]->(v2)

expresses an arbitrary edge between two arbitrary vertexes v1,v2 but does not assign a name to the edge

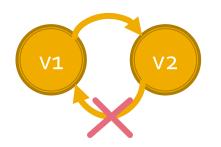


 These are called anonymous vertexes and edges

- A basic unit (an edge between two vertexes) can be negated to indicate that the edge should not be present in the graph
  - (v1)-[]->(v2); !(v2)-[]->(v1)

means

Edges from v1 to v2 but no edges from v2 to v1



- The find(motif) method of GraphFrame is used to select motifs
  - motif
    - DSL representation of the structural pattern

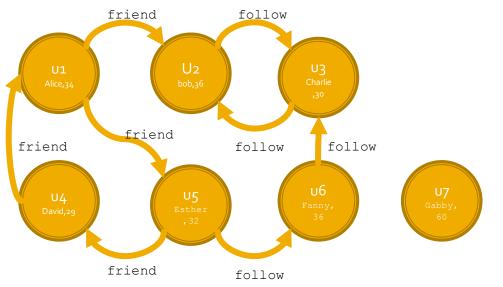
- find() returns a DataFrame of all the paths matching the structural motif/pattern
  - One path per record
  - The returned DataFrame will have a column for each of the named elements (vertexes and edges) in the structural pattern/motif
    - Each column is a struct
      - The fields of each struct are the labels/features of the associated vertex or edge
  - It can return duplicate rows/records
    - If there are many paths connecting the same nodes

- More complex queries on the structure and content of the patterns can be expressed by applying filters to the result DataFrame
  - i.e., more complex queries can be applied by combing find() and filter()

 Find the paths/subgraphs matching the pattern

(v1) - [e1] -> (v2); (v2) - [e2] -> (v1)

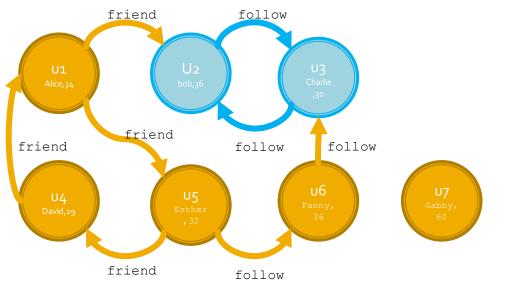
Store the result in a DataFrame



 Find the paths/subgraphs matching the pattern

(v1) - [e1] -> (v2); (v2) - [e2] -> (v1)

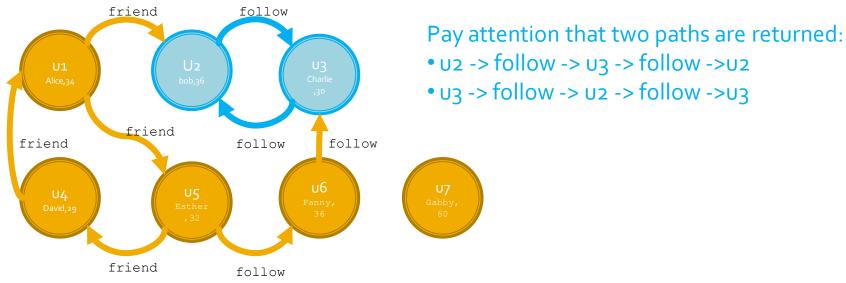
Store the result in a DataFrame



 Find the paths/subgraphs matching the pattern

(v1) - [e1] -> (v2); (v2) - [e2] -> (v1)

Store the result in a DataFrame



 Find the paths/subgraphs matching the pattern

(v1) - [e1] -> (v2); (v2) - [e2] -> (v1)

Content of the returned DataFrame

++	e1	V2	62	I
[u2, Bob, 36]	[u2, u3, follow]	[u3, Charlie, 30]	[U3, U2, follow]	
[u3, Charlie, 30]	[u3, u2, follow]	[u2, Bob, 36]	[U2, U3, follow]	

 Find the paths/subgraphs matching the pattern

(v1) - [e1] -> (v2); (v2) - [e2] -> (v1)
 Content of the returned DataFrame

+	+		+		+		+
V1	į [	e1		V2	İ	e2	
+	+		+		+		+

There is one column for each (distinct) named vertex and edge of the structural pattern

 Find the paths/subgraphs matching the pattern

(v1) - [e1] -> (v2); (v2) - [e2] -> (v1)

Content of the returned DataFrame

V1	I	V2	e2	
[u2, Bob, 36]	[u2, u3, follow]		כ]  [טȝ, ט₂, follow]	
[uȝ, Charlie, ȝo	]  [u3, u2, follow]	[u2, Bob, 36]	[U2, U3, follow]	

The records are associated with the vertexes and edges of the selected paths

 Find the paths/subgraphs matching the pattern

(v1) - [e1] -> (v2); (v2) - [e2] -> (v1)

Content of the returned DataFrame

+	V1	eı	-+   -+	V2	e2	-+   
		[U2, U3, follow] [U3, U2, follow]			[U3, U2, follow] [U2, U3, follow]	

All columns are associated with the data type "struct". Each struct has the same "schema/features" of the associated vertex or edge.

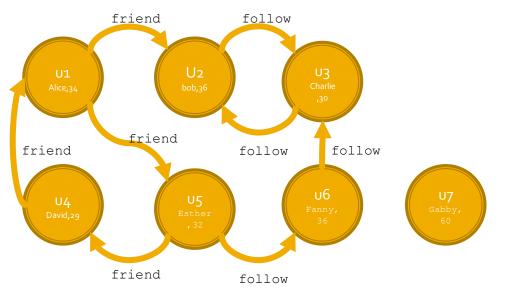
from graphframes import GraphFrame

# Create the graph
g = GraphFrame(v, e)

# Retrieve the motifs associated with the pattern # vertex -> edge -> vertex -> edge ->vertex motifs = g.find("(v1)-[e1]->(v2); (v2)-[e2]->(v1)")

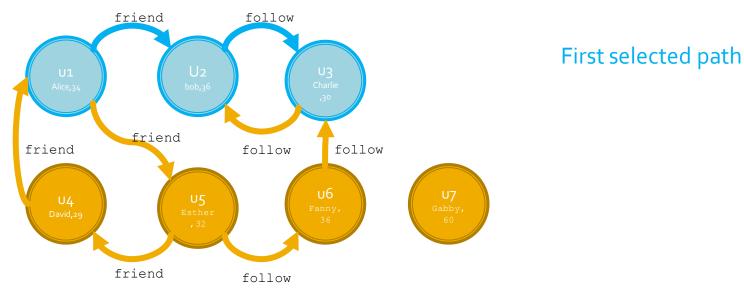
 Find the paths/subgraphs matching the pattern

(v1)- [friend] -> (v2); (v2)- [follow] -> (v3)
Store the result in a DataFrame



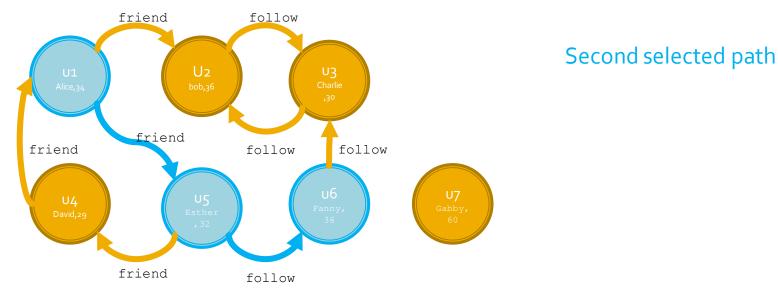
 Find the paths/subgraphs matching the pattern

(v1)- [friend] -> (v2); (v2)- [follow] -> (v3)
 Store the result in a DataFrame



 Find the paths/subgraphs matching the pattern

(v1)- [friend] -> (v2); (v2)- [follow] -> (v3)
Store the result in a DataFrame



from graphframes import GraphFrame

# Create the graph
g = GraphFrame(v, e)

# Retrieve the motifs associated with the pattern
# vertex -> edge -> vertex -> edge ->vertex
motifs = g.find("(v1)-[friend]->(v2); (v2)-[follow]->(v3)")

# Filter the motifs (the content of the motifs DataFrame)
# Select only the ones matching the pattern
# vertex -> friend-> vertex -> follow ->vertex
motifsFriendFollow = motifs\
.filter("friend.relationship='friend'AND follow.relationship='follow'")

# Retrieve the motifs associated with the pattern
# vertex -> edge -> vertex -> edge ->vertex
motifs = g.find("(v1)-[friend]->(v2); (v2)-[follow]->(v3)")

# Filter the motifs (the content of the motifs DataFrame)
# Select only the ones matching the pattern
# vertex -> friend-> vertex -> follow ->vertex
motifsFriendFollow = motifs\
.filter("friend.relationship='friend'AND follow.relationship='follow'")

Columns friend and follow are structs with three fields/attributes

- src
- dst
- relationship

# Motif finding: Example 2

# Retrieve the motifs associated with the pattern
# vertex -> edge -> vertex -> edge ->vertex
motifs = g.find("(v1)-[friend]->(v2); (v2)-[follow]->(v3)")

# Filter the motifs (the content of the motifs DataFrame)
# Select only the ones matching the pattern
# vertex -> friend-> vertex -> follow ->vertex
motifsFriendFollow = motifs\
.filter("friend.relationship ='friend'AND follow.relationship='follow'")

To access a field of a struct column use the syntax columnName.field

#### **Basic statistics**

- Some specific properties are provided to compute basic statistics on the degrees of the vertexes
  - degrees
  - inDegrees
  - outDegrees
- The returned result of each of this property is a DataFrame with
  - id
  - (in/out)Degree value

## **Basic statistics: degrees**

#### degrees

- Returns the degree of each vertex
  - i.e., the number of edges associated with each vertex
- The result is stored in a DataFrame with Columns (vertex) "id" and "degree"
  - One record per vertex
  - Only the vertexes with degree>=1 are stored in the returned DataFrame

## **Basic statistics: inDegrees**

#### inDegrees

- Returns the in-degree of each vertex
  - i.e., the number of in-edges associated with each vertex
- The result is stored in a DataFrame with Columns (vertex) "id" and "inDegree"
  - One record per vertex
  - Only the vertexes with in-degree>=1 are stored in the returned DataFrame

## **Basic statistics: outDegrees**

#### outDegrees

- Returns the out-degree of each vertex
  - i.e., the number of out-edges associated with each vertex
- The result is stored in a DataFrame with Columns (vertex) "id" and "outDegree"
  - One record per vertex
  - Only the vertexes with out-degree>=1 are stored in the returned DataFrame

- Given the input graph, compute
  - Degree of each vertex
  - inDegree of each vertex
  - outDegree of each vertex

from graphframes import GraphFrame

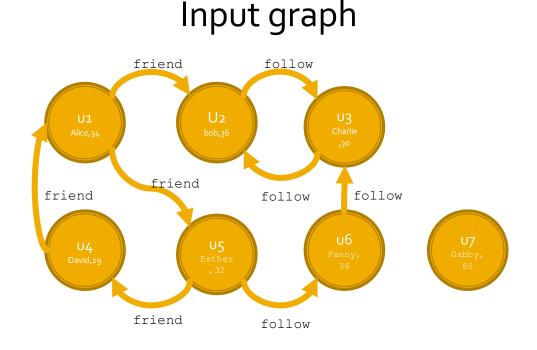
# Create the graph
g = GraphFrame(v, e)

# Retrieve the DataFrame with the information about the degree of # each vertex vertexesDegreesDF = g.degrees

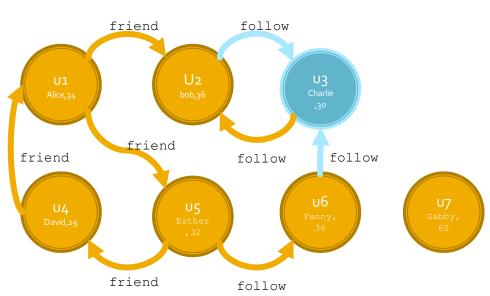
# Retrieve the DataFrame with the information about the in-degree of # each vertex vertexesInDegreesDF = g.inDegrees

# Retrieve the DataFrame with the information about the out-degree of # each vertex vertexesOutDegreesDF = g.outDegrees

 Given the input graph, select only the ids of the vertexes with at least 2 in-edges

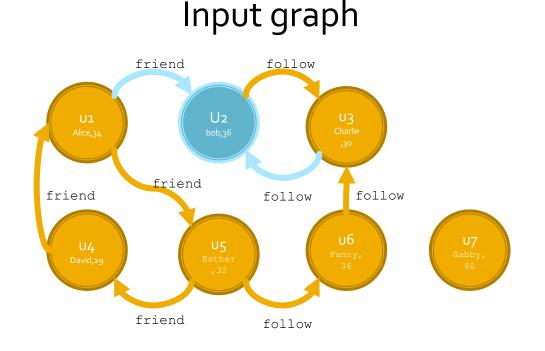


 Given the input graph, select only the ids of the vertexes with at least 2 in-edges

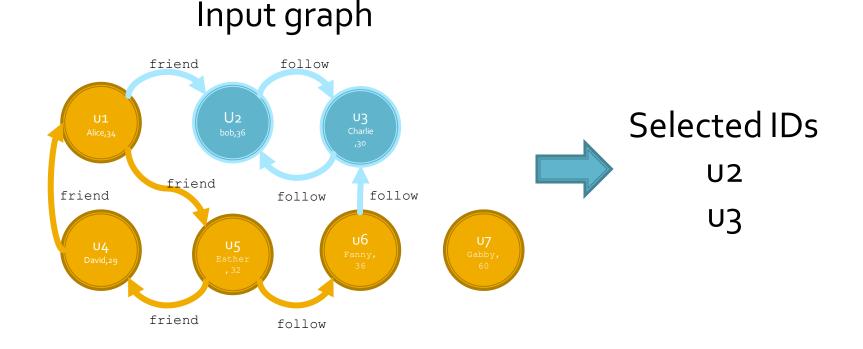


Input graph

 Given the input graph, select only the ids of the vertexes with at least 2 in-edges



 Given the input graph, select only the ids of the vertexes with at least 2 in-edges



from graphframes import GraphFrame

# Create the graph
g = GraphFrame(v, e)

# Retrieve the DataFrame with the information about the in-degree of # each vertex vertexesInDegreesDF = g.inDegrees

# Select only the vertexes with and in-degree value >=2
selectedVertexesDF = vertexesInDegreesDF.filter("inDegree>=2")

# Select only the content of Column id
selectedVertexesIDsDF = selectedVertexesDF.select("id")