#### GRAPHX

#### APACHE SPARK

## Spark comes with some additional packages that make it truly general - purpose

Spark Core

Storage System Cluster
manager

#### APACHE SPARK

Spark SQL

Spark Streaming

MLlib GraphX

Spark Core

Storage System

Cluster manager

#### APACHE SPARK

MLlib GraphX

GraphX is a library for graph algorithms

#### APACHE SPARK

MLlib GraphX

Many interesting datasets can be represented as graphs

#### APACHE SPARK

MLlib GraphX

Social networks, linked webpages etc

#### APACHE SPARK

MLlib GraphX

With GraphX you can represent and then perform computations across these datasets

#### APACHE SPARK

Just like with MLlib, GraphX abstracts the programmer from having to implement Graph algorithms

GraphX

#### APACHE SPARK

Because of RDDs and in-memory computation, GraphX on Spark gives you better performance than other computing frameworks (like MapReduce)

GraphX

GraphX

## GraphX is only available for Scala

For Python, a separate module called GraphFrames is available

### GraphFrames

# This module has most of the functionality that GraphX provides

GraphX

It is slated to be integrated into Spark and with GraphX in the near future

### GraphFrames

GraphX

## Let's use GraphFrames to explore an interesting dataset

### Marvel Social Network

GraphX



GraphX

### The Marvel Comic Book Universe has thousands of characters



GraphX

The relationships between these characters could be explored using a graph

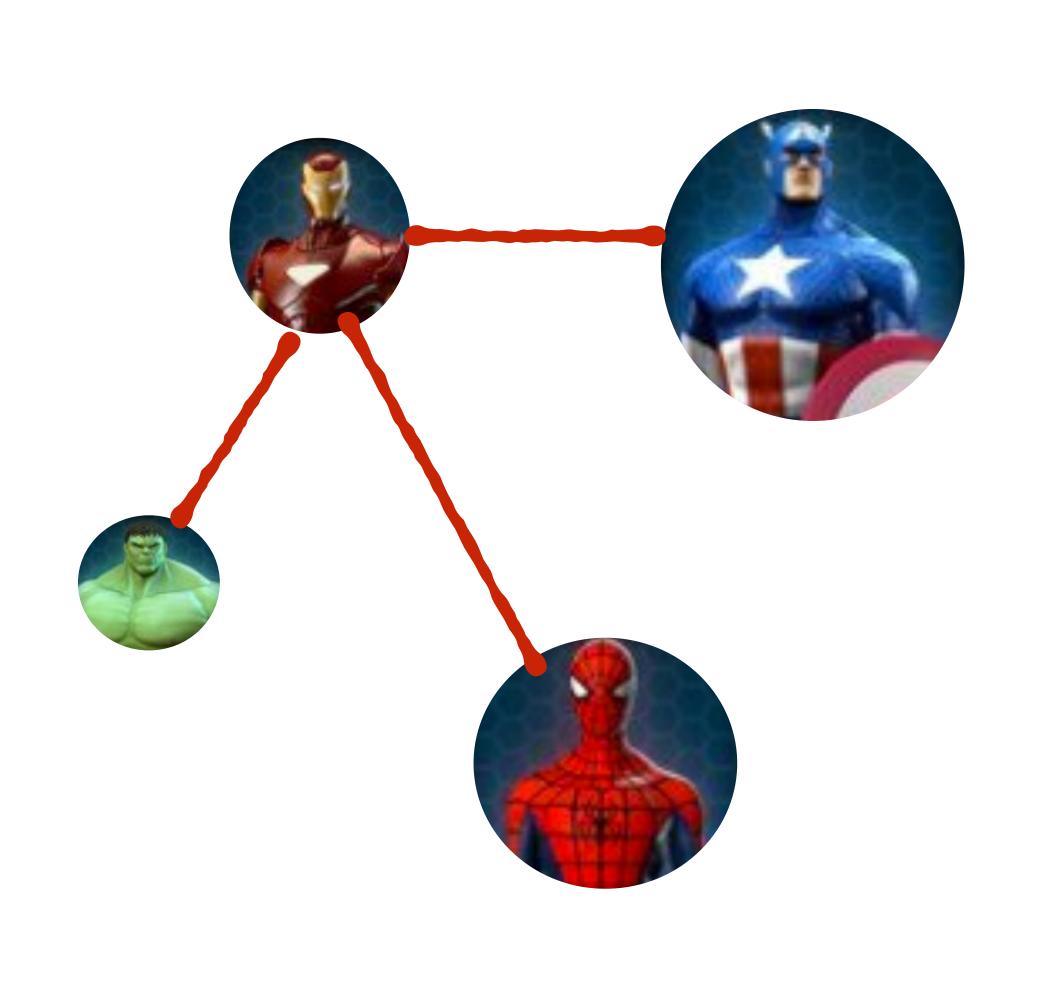


GraphX

#### The graph would have

1. Characters as vertices

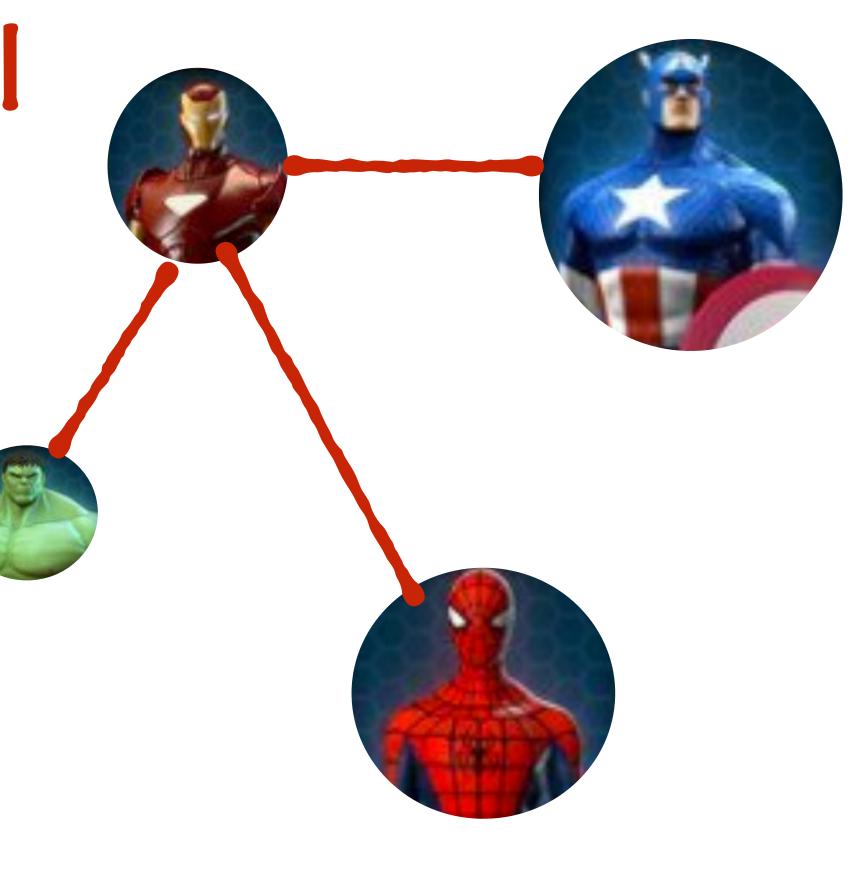
2. An edge representing if 2 characters appear together in a book



GraphX

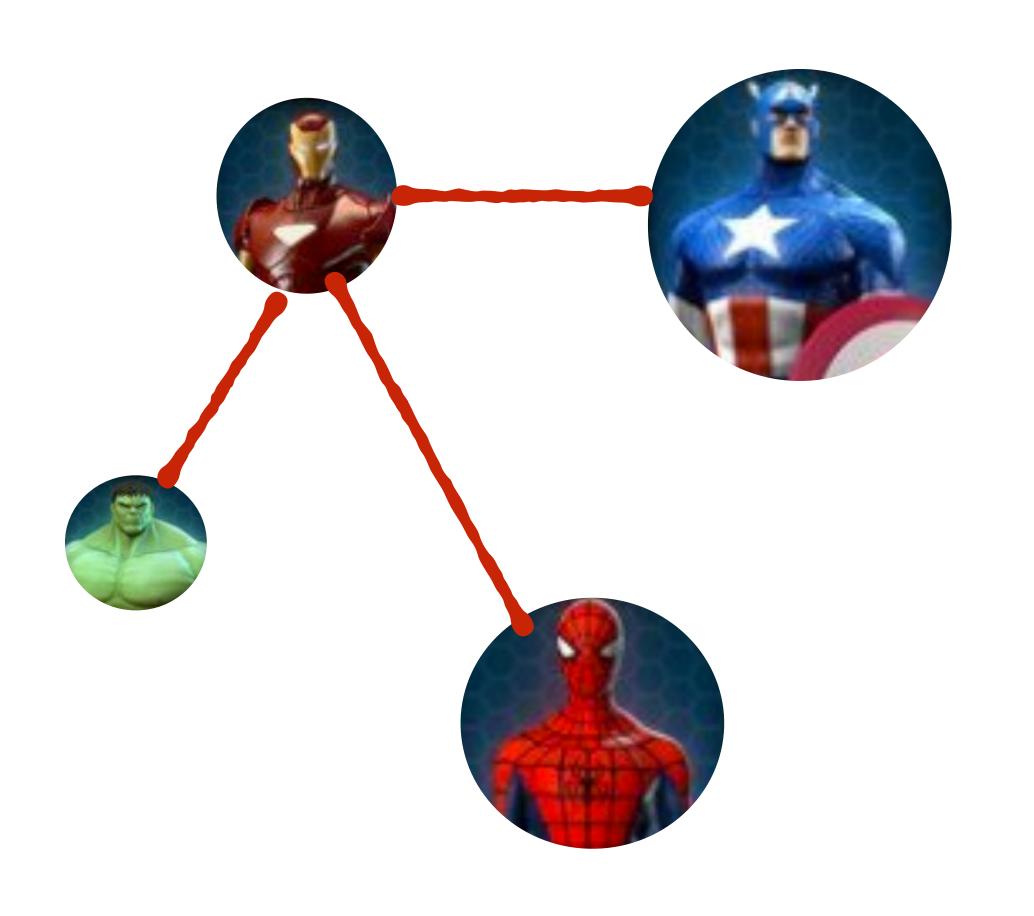
1. Size of the vertices is proportional to number of times the character appears in any comic book

2. Edge weight is proportional to the number of times the characters appear together



GraphX

Such graphs are called Co-occurence networks



GraphX

## The dataset for this exercise was constructed using raw data provided by

<u>Author ©:Joe Miro</u>

#### who did a study on

Social characteristics of the Marvel Universe

GraphX

#### The dataset has

A Vertices file

Character id, Name, # Times the character appears in any book

An Edges file

Character 1, Character 2, # Times they appear together

GraphX

## To represent a graph in Spark, you need

A Vertices Pataframe

An Edges Pataframe

GraphX

A Vertices PataFrame An Edges PataFrame

Recall: A PataFrame is like an in-memory table

GraphX

#### A Vertices Pataframe

# This PataFrame must have an "id" field

GraphX

A Vertices PataFrame "id" field

It can have other fields as well, but these are left to the user's discretion

GraphX

A Vertices Pataframe "id" field The other fields may represent attributes of the vertex Ex: Name

GraphX

A Vertices PataFrame "io" te

#### An Edges Vataframe

# This must have an "src" field and a "dst" field

GraphX

A Vertices PataFrame "10" tie 0

An Edges Pataframe It can have other fields corresponding to edge attributes Ex: Edge weight

GraphX

from pyspark.sql import SQLContext, Row
sqlC=SQLContext(sc)

# We need an SQLContext to set up PataFrames

```
edges=sqlC.createDataFrame\
   (sc.textFile(edgesPath)\
   .map(lambda x:x.split(","))\
   .map(lambda x:[int(y) for y in x])\
   ,["src","dst","wt"])
```

### GraphX

### Here is the edges VataFrame

```
GraphX
```

```
edges=sqlC.createDataFrame
  (sc.textFile(edgesPath)\
  .map(lambda x:x.split(","))\
  .map(lambda x:[int(y) for y in x])\
  ,["src","dst","wt"])
```

## We've seen an example where we used sqlC.read to create a DataFrame

```
GraphX
```

```
edges=sqlC.createDataFrame
  (sc.textFile(edgesPath))
  .map(lambda x:x.split(",")))
  .map(lambda x:[int(y) for y in x]))
  ,["src","dst","wt"])
```

## The read method is used when you are reading data directly from file

```
GraphX
```

```
edges=sqlC.createDataFrame
  (sc.textFile(edgesPath))
  .map(lambda x:x.split(",")))
  .map(lambda x:[int(y) for y in x]))
  ,["src","dst","wt"])
```

# This is another way to create a DataFrame

```
edges=sqlC.createDataFrame
  (sc.textFile(edgesPath)\
  .map(lambda x:x.split(","))\
  .map(lambda x:[int(y) for y in x])\
  ,["src","dst","wt"])
```

GraphX

# We need to give this method an RDD

```
GraphX
```

```
edges=sqlC.createDataFrame\
  (sc.textFile(edgesPath)\
  .map(lambda x:x.split(","))\
  .map(lambda x:[int(y) for y in x])\
  ,["src","dst","wt"])
```

# We need to give this method an RPP

```
edges=sqlC.createDataFrame\
  (sc.textFile(edgesPath)\
  .map(lambda x:x.split(","))\
  .map(lambda x:[int(y) for y in x])\
  ,["src","dst","wt"])
```

### GraphX

### And a schema

```
GraphX
```

```
edges=sqlC.createDataFrame\
  (sc.textFile(edgesPath)\
  .map(lambda x:x.split(","))\
  .map(lambda x:[int(y) for y in x])\
  ,["src","dst","wt"])
```

## We know that our edges PataFrame must have "src" and "dst" fields

```
GraphX
```

```
edges=sqlC.createDataFrame\
  (sc.textFile(edgesPath)\
  .map(lambda x:x.split(","))\
  .map(lambda x:[int(y) for y in x])\
  ,["src","dst","wt"])
```

# Using createDataFrame we can explicitly specify the schema for the DataFrame

```
GraphX
```

```
edges=sqlC.createDataFrame\
  (sc.textFile(edgesPath)\
  .map(lambda x:x.split(","))\
  .map(lambda x:[int(y) for y in x])\
  ,["src","dst","wt"])
```

### In addition to the mandatory fields, we also specify a weight

```
edges=sqlC.createDataFrame\
   (sc.textFile(edgesPath)\
   .map(lambda x:x.split(","))\
   .map(lambda x:[int(y) for y in x])\
   ,["src","dst","wt"])
```

GraphX

### This is the count of the number of times the 2 characters appear together

```
GraphX
```

```
edges=sqlC.createDataFrame\
  (sc.textFile(edgesPath)\
  .map(lambda x:x.split(","))\
  .map(lambda x:[int(y) for y in x])\
  ,["src","dst","wt"])
```

Load the data from file into an RDD, parse the rows and give it to createDataFrame

GraphX

```
vertices=sqlC.createDataFrame\
   (sc.textFile(verticesPath)\
   .map(lambda x:x.split("|"))\
   .map(lambda x:[int(x[0]),x[1],int(x[2])])\
   ["id","name","wt"]
```

#### Similarly, we can set up a vertices DataFrame with a schema

GraphX

```
vertices=sqlC.createDataFrame\
   (sc.textFile(verticesPath)\
   .map(lambda x:x.split("|"))\
   .map(lambda x:[int(x[0]),x[1],int(x[2])])\
   ,["id","name","wt"])
```

### The vertices Pataframe needs an "id" field

GraphX

```
vertices=sqlC.createDataFrame\
    (sc.textFile(verticesPath)\
    .map(lambda x:x.split("|"))\
    .map(lambda x:[int(x[0]),x[1],int(x[2])])\
    ,["id", name","wt"])
```

### These 2 are attributes of the vertices

GraphX

```
vertices=sqlC.createDataFrame\
    (sc.textFile(verticesPath)\
    .map(lambda x:x.split("|"))\
    .map(lambda x:[int(x[0]),x[1],int(x[2])])\
    ,["id", name","wt"])
```

The name of the character and the number of times he appears in the books

GraphX

```
vertices=sqlC.createDataFrame\
    (sc.textFile(verticesPath)\
    .map(lambda x:x.split("|"))\
    .map(lambda x:[int(x[0]),x[1],int(x[2])])\
    ,["id","name","wt"])
```

Load the data from file into an RDD, parse the rows and give it to createDataFrame

GraphX

## Now we can create a GraphFrame

```
from graphframes import *
marvelSocial=GraphFrame vertices, edges)
```

GraphX

#### GraphFrame

#### has a bunch of interesting methods and attributes

GraphX

#### GraphFrame

# The degrees attribute will tell us the number of connections a vertex has

GraphX

#### GraphFrame

Use the degrees attribute to find the most important (well-connected) characters

GraphX

degrees=marvelSocial.degrees

```
degrees.take(10)
```

```
Row(id=2031, degree=822),
Row(id=3831, degree=38),
Row(id=3231, degree=648),
Row(id=3031, degree=352),
Row(id=1831, degree=120),
Row(id=2831, degree=188),
Row(id=2631, degree=76),
Row(id=3431, degree=46),
```

### degrees is another PataFrame

GraphX

degrees=marvelSocial.degrees

degrees.take(10)

```
Row(id=2031, degree=822),
Row(id=3831, degree=38),
Row(id=3231, degree=648),
Row(id=3031, degree=352),
Row(id=1831, degree=120),
Row(id=2831, degree=188),
Row(id=2631, degree=76),
Row(id=3431, degree=46),
```

### The character id

GraphX

degrees=marvelSocial.degrees

degrees.take(10)

```
[Row(id=2031, degree=822),
Row(id=3831, degree=38),
Row(id=3231, degree=648),
Row(id=3031, degree=352),
Row(id=1831, degree=120),
Row(id=2831, degree=188),
Row(id=2631, degree=76),
Row(id=3431, degree=46),
```

# The number of connections for that character

GraphX

degrees=marvelSocial.degrees

```
degrees.take(10)
```

```
[Row(id=2031, degree=822),
Row(id=3831, degree=38),
Row(id=3231, degree=648),
Row(id=3031, degree=352),
Row(id=1831, degree=120),
Row(id=2831, degree=188),
Row(id=2631, degree=76),
Row(id=3431, degree=46),
```

We can sort the DataFrame by the degree column to get the Top 10 characters

GraphX

```
top10=degrees sort(col("degree").desc()).take(10)
```

### DataFrames have a sort method

GraphX

```
top10=degrees sort(col("degree").desc()].take(10)
```

You can specify a column name and sort them by that column

GraphX

```
top10=degrees.sort(col("degree").desc()).take(10)
```

Here we sort in descending order of the degree column

GraphX

```
top10=degrees.sort(col("degree").desc()).take(10)
```

We can print out the character names for these ids

GraphX

# Each member of the topl 0 list is a Row Object

GraphX

```
for character in top10:
    charId=character.id
    print vertices.filter("id="+str(charId)).first().name
```

### Extract the character id from the Row

GraphX

```
for character in top10:
    charId=character.id
    print vertices.filter("id="+str(charId)).first().name
```

#### vertices is a PataFrame

GraphX

```
for character in top10:
    charId=character.id
    print vertices.filter("id="+str(charId)).first().name
```

#### The filter transformation for PataFrames is slightly different

GraphX

```
for character in top10:
    charId=character.id
    print vertices.filter("id="+str(charId)).first().name
```

# This transformation uses a string as a condition instead of a lambda function

GraphX

```
for character in top10:
    charId=character.id
    print vertices.filter("id="+str(charId)).first().name
```

## The string uses columns in the DataFrame directly

GraphX

```
for character in top10:
    charId=character.id
    print vertices.filter("id="+str(charId)) .first().name
```

### This condition will filter the vertex corresponding to the character id

GraphX

```
for character in top10:
    charId=character.id
    print vertices.filter("id="+str(charId)).first().name
```

## Extract the first row which matches and print the name

GraphX

```
for character in top10:
    charId=character.id
    print vertices.filter("id="+str(charId)).first().name

These are the top10:
    CharId=character.id
    print vertices.filter("id="+str(charId)).first().name
```

CAPTAIN AMERICA
SPIDER-MAN/PETER PARKER
IRON MAN/TONY STARK
THING/BENJAMIN J. GRIMM
WOLVERINE/LOGAN
MR. FANTASTIC/REED RICHARDS
HUMAN TORCH/JOHNNY STORM
SCARLET WITCH/WANDA MAXIMOFF
THOR/DR. DONALD BLAKE/SIGURD JARLSON II/JAKE OLSON/LOREN OLSON

BEAST/HENRY &HANK& P. MCCOY

# These are the top 10 characters in the Marvel Social Network

GraphX

# The module has several interesting methods

for implementing Graph Algorithms

#### Graph Algorithms Breadth first Search PageRank Connected Components Triangle search

GraphX

#### Graph Algorithms

Breadth first Search

PageRank

Connected Components

Triangle search

GraphX

These are a few of the standard Graph processing algorithms available as built-in methods