

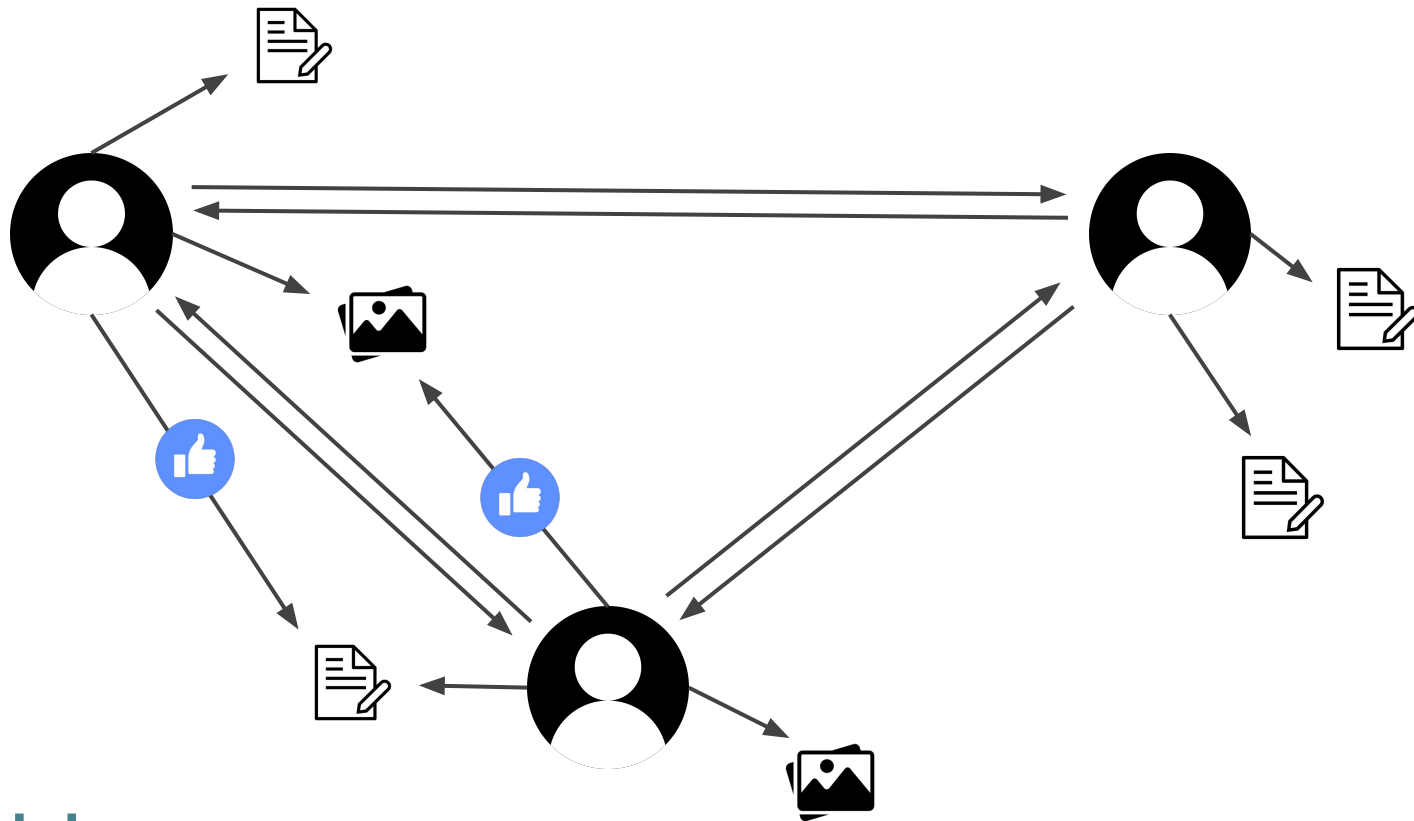
# TP1 - BCC 362

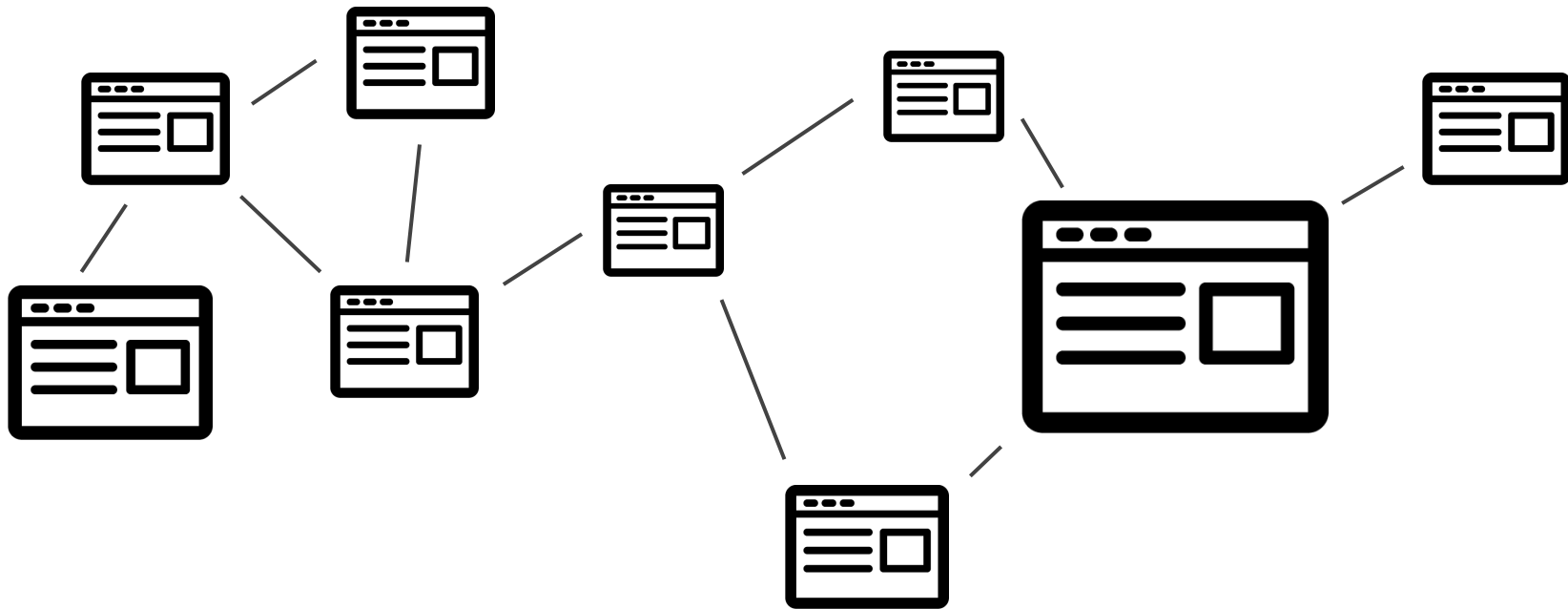


- Graphx | Apache Spark
- É um componente para computação paralela de grafos. Trata-se de uma estrutura de **processamento de grafos distribuídos** que roda sobre o Spark.

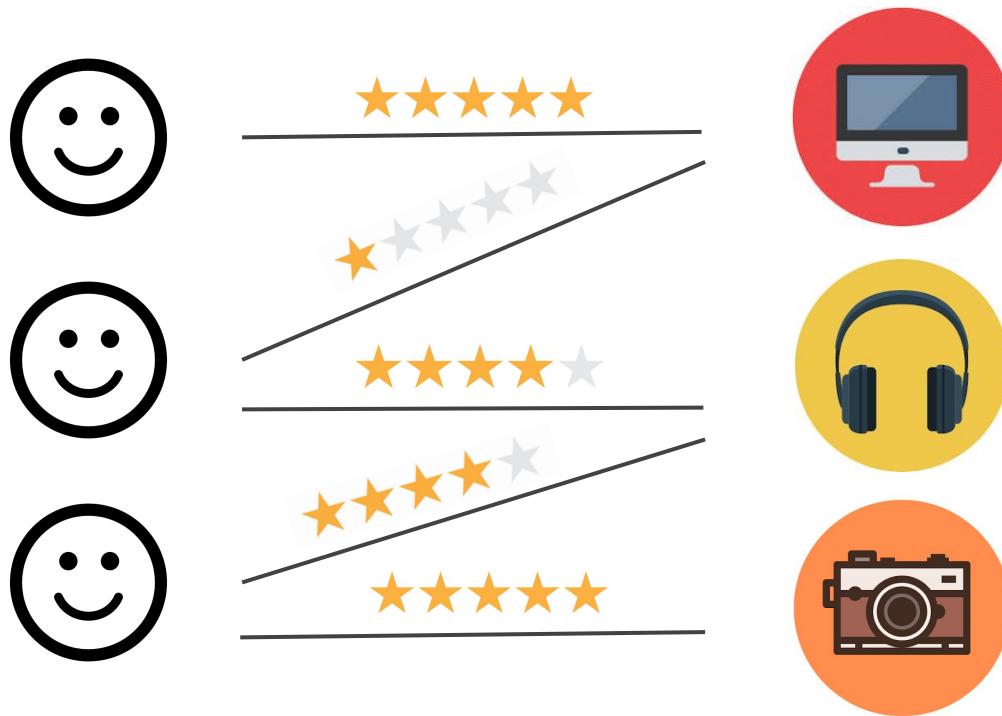
# GRAFOS







Web graphs

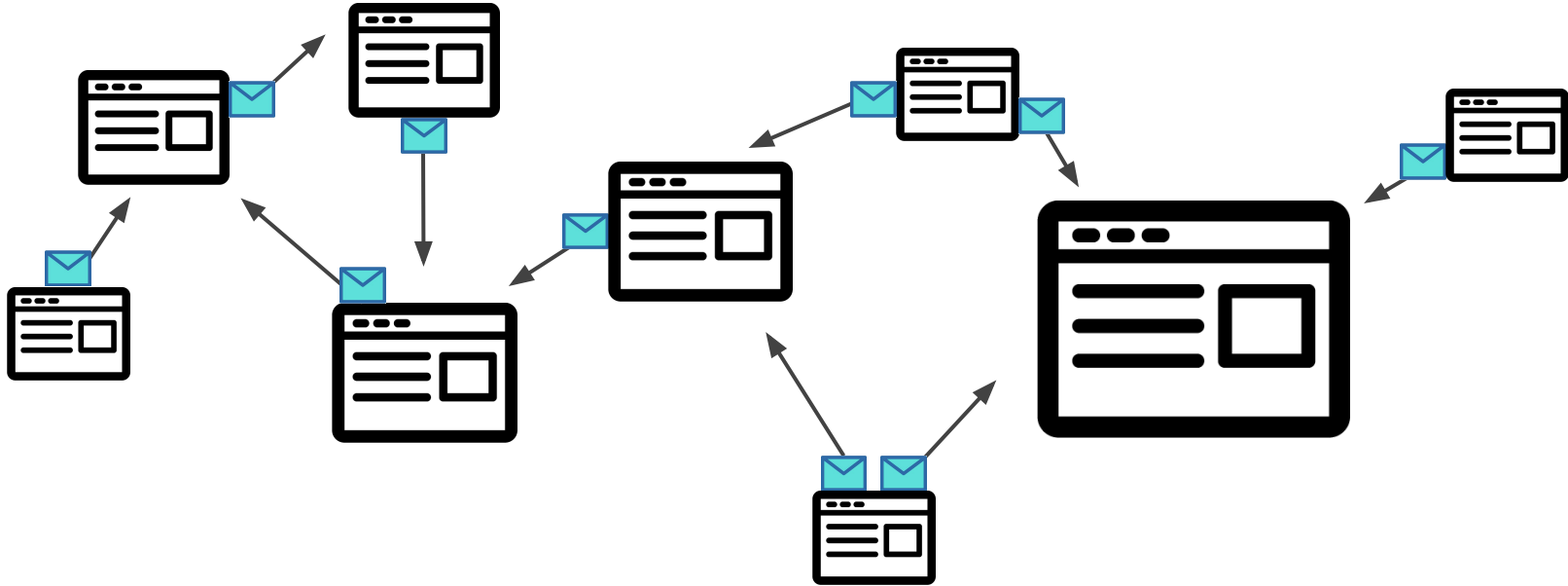


Relação item-usuário

# ALGORITMOS PARA GRAFOS

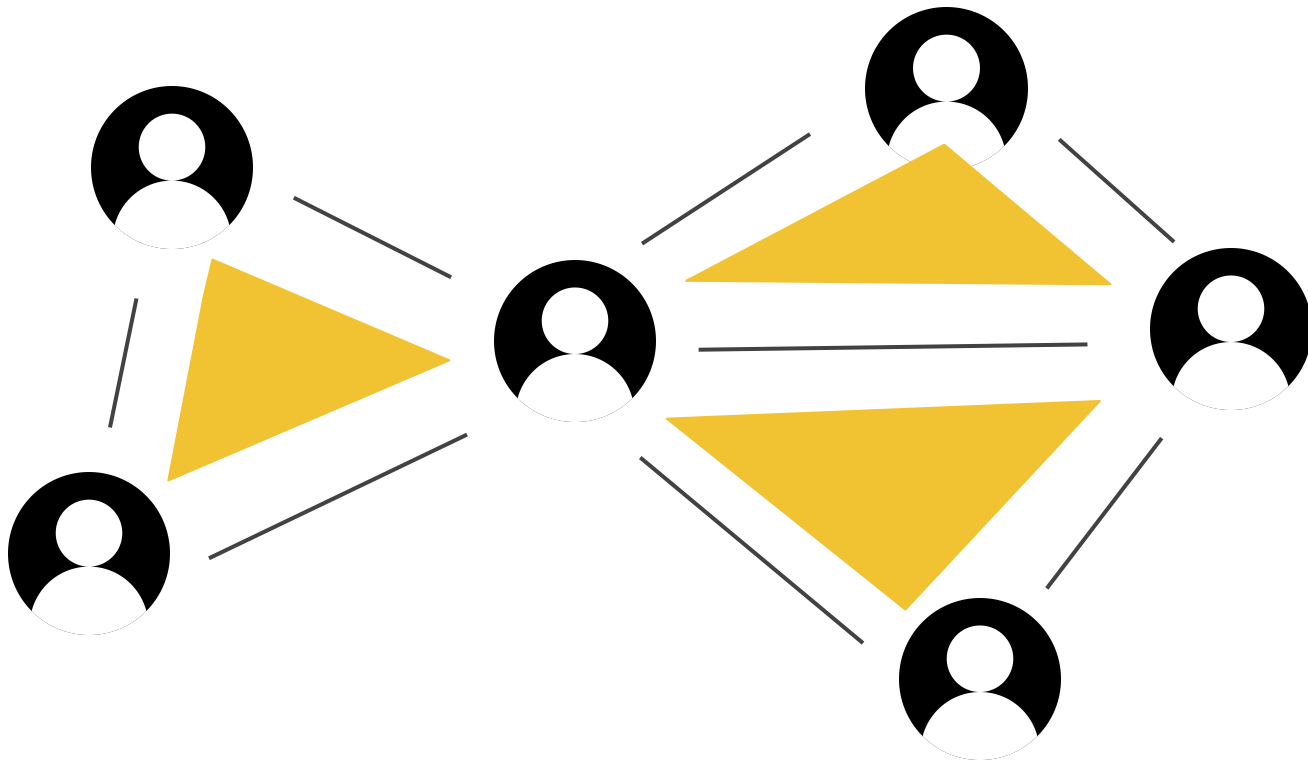


# ALGORITMOS PARA GRAFOS



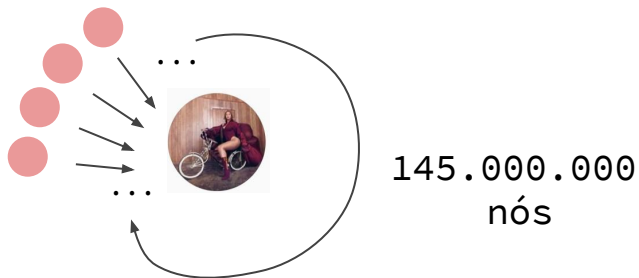
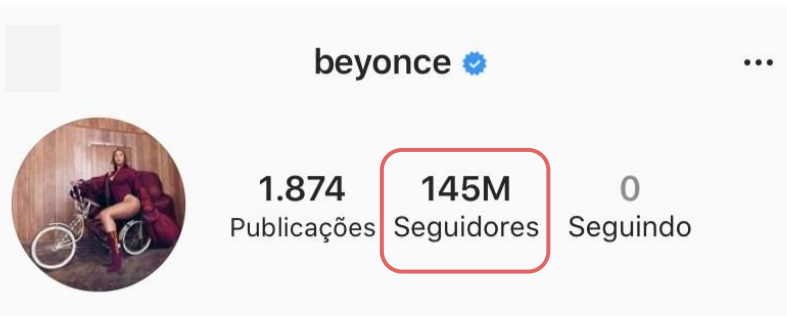
PageRank





## Triangle Counting

## Vértices com grau alto

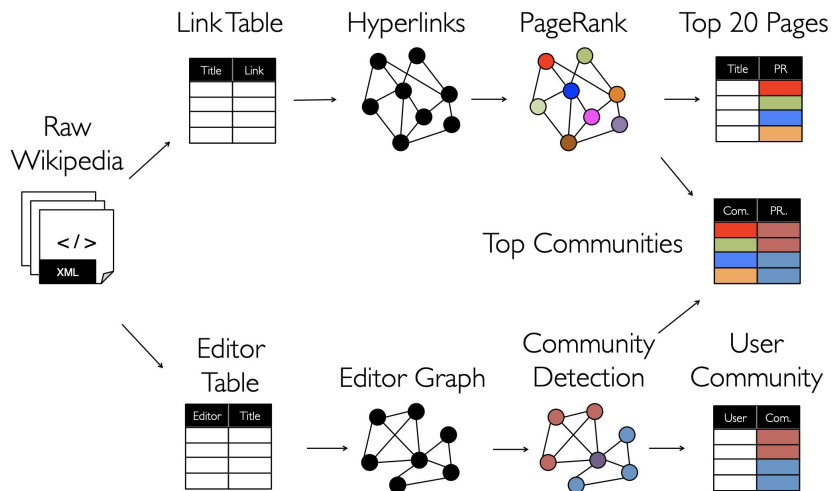


## DESAFIOS

**ARMAZENAMENTO:** como armazenar grafos quando todos seus nós não cabem em uma máquina só?

**API:** como expor vértices vizinhos ao paralelismo?

## Pipelines complexos



## SOLUÇÃO

Incorporar um sistema orientado a tabelas (SPARK) ao processamento de grafos

**ARMAZENAMENTO:** como armazenar grafos como tabelas?

**COMPUTAÇÃO:** como expressar operações em grafos como operações em tabelas (map, reduce, join, etc)?

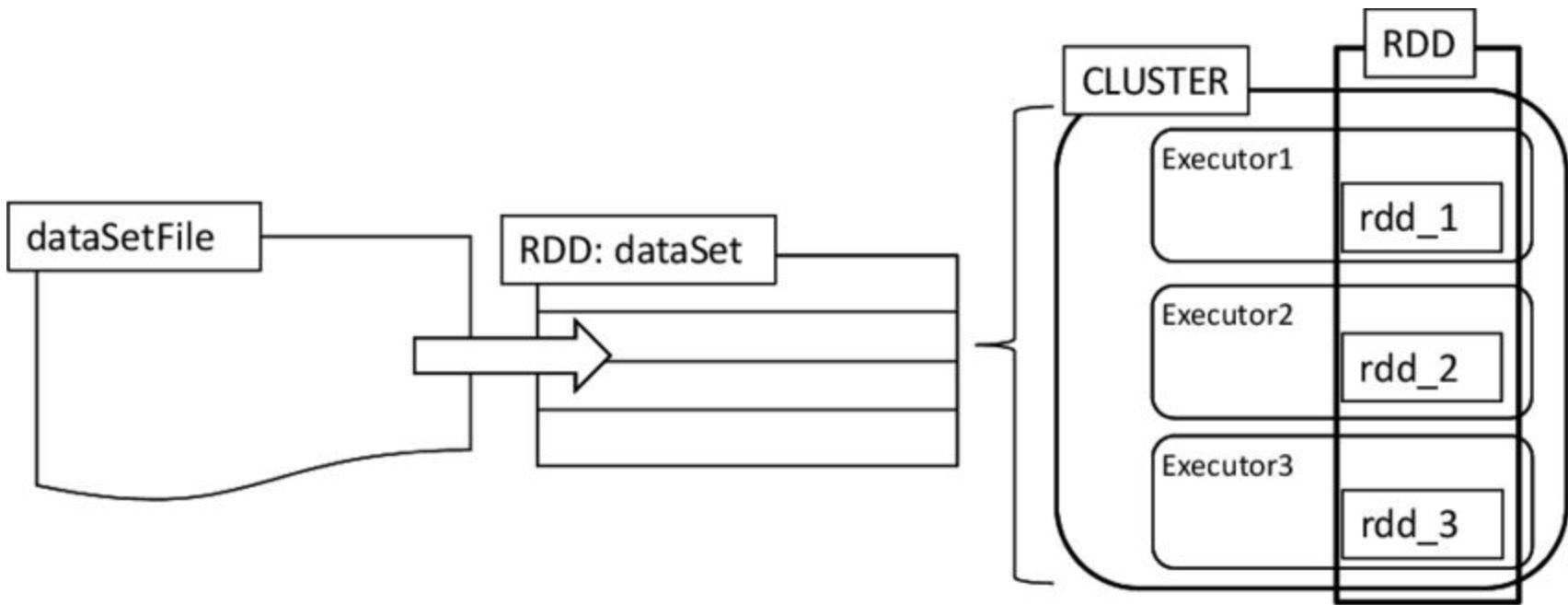
**API:** como apresentar essas abordagens ao usuário?

## DESAFIOS

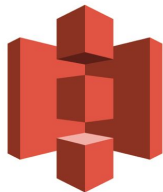
# SPARK



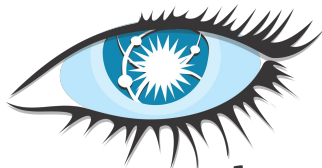
# RESILIENT DISTRIBUTED DATASETS (RDDs)



# RDDs E DFS



Amazon S3



*cassandra*

# SETUP BÁSICO



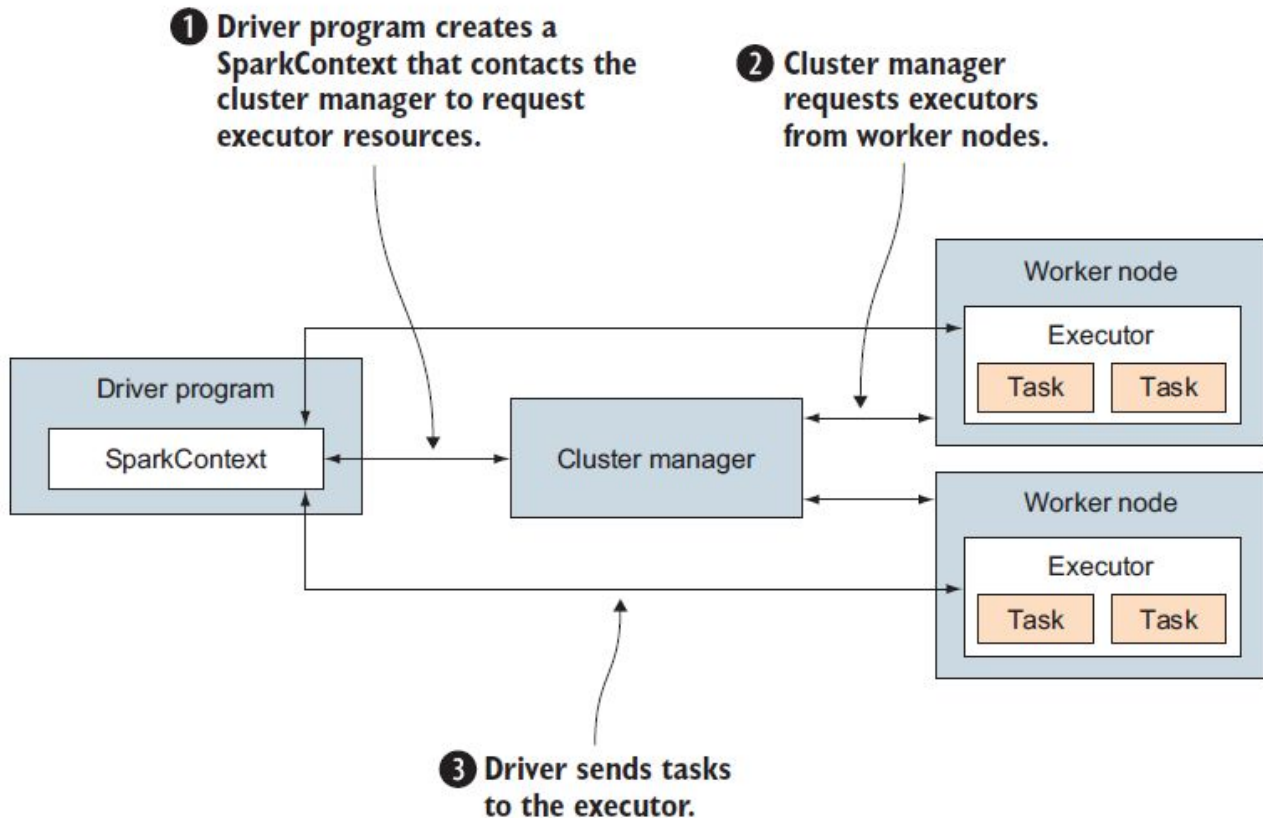
## Distributed Storage

- HDFS
- Cassandra
- S3
- Other cloud storage vendors

## Cluster Manager

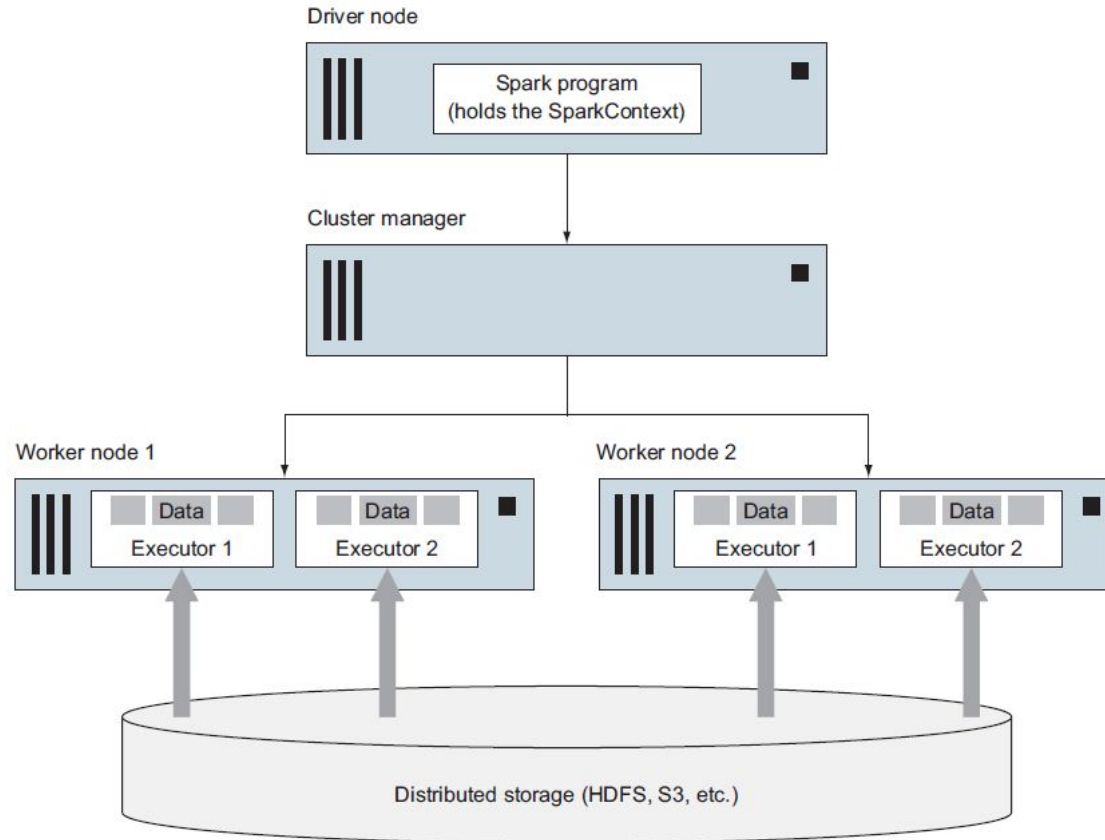
- Standalone
- YARN
- Mesos

# FUNCIONAMIENTO BÁSICO





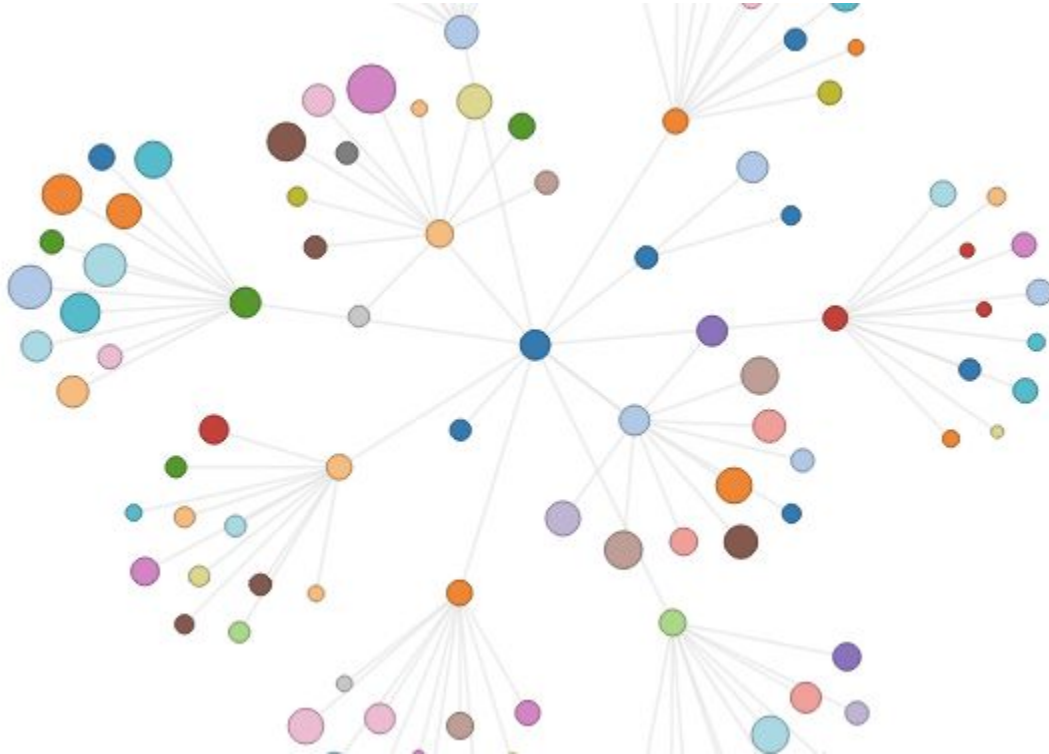
# FUNCIONAMIENTO BÁSICO



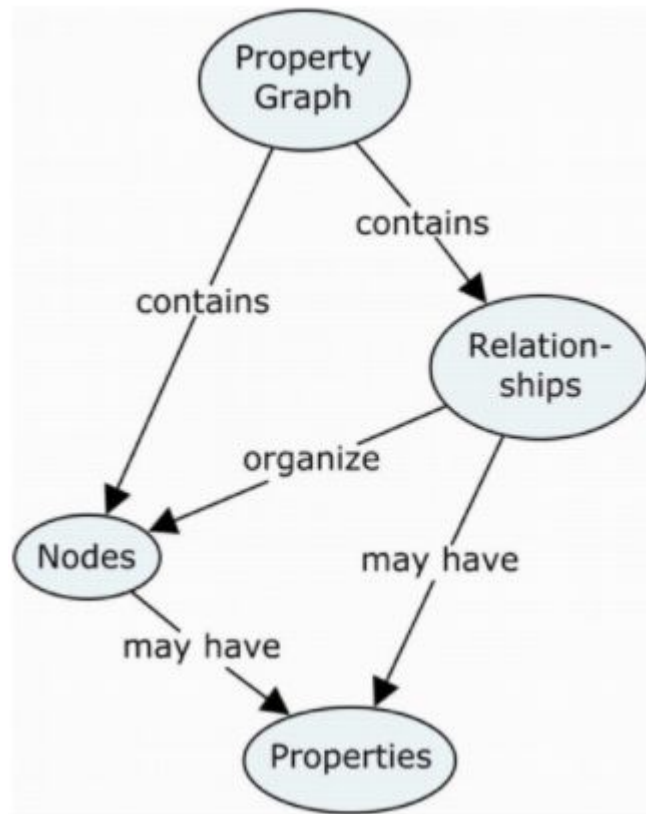
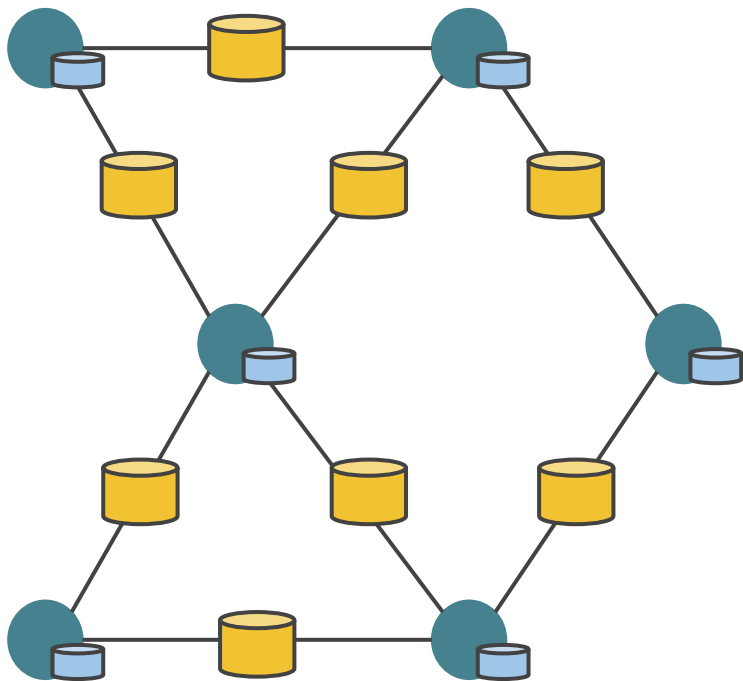
# GRAPHX



# PROPERTY GRAPHS



# PROPERTY GRAPHS



# PROPERTY GRAPHS

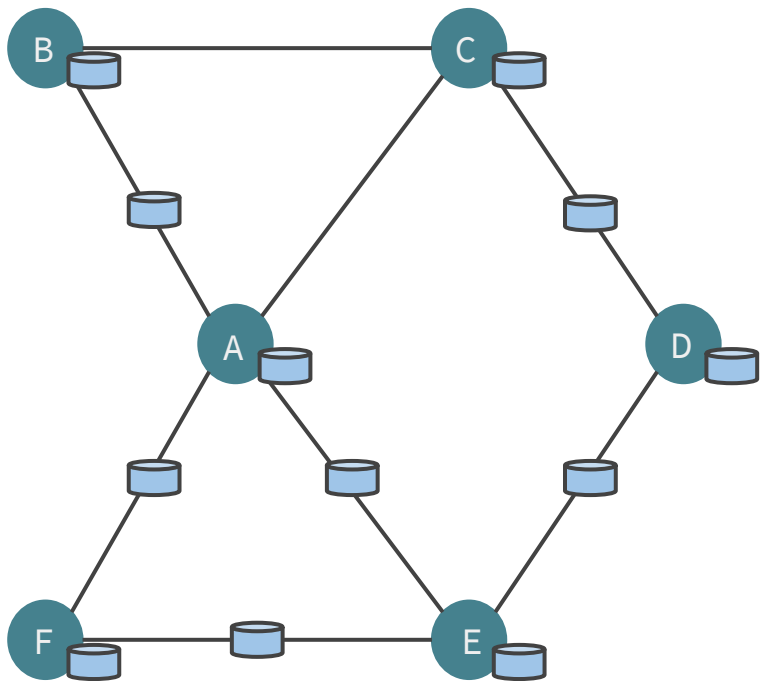


Tabela de vértices  
RDD

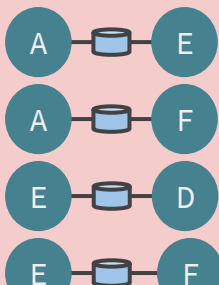
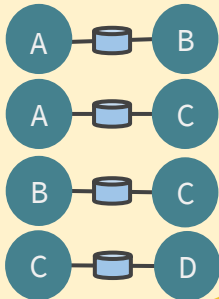


máquina 1



máquina 2

Tabela de arestas  
RDD



# PROS / CONS



GraphX

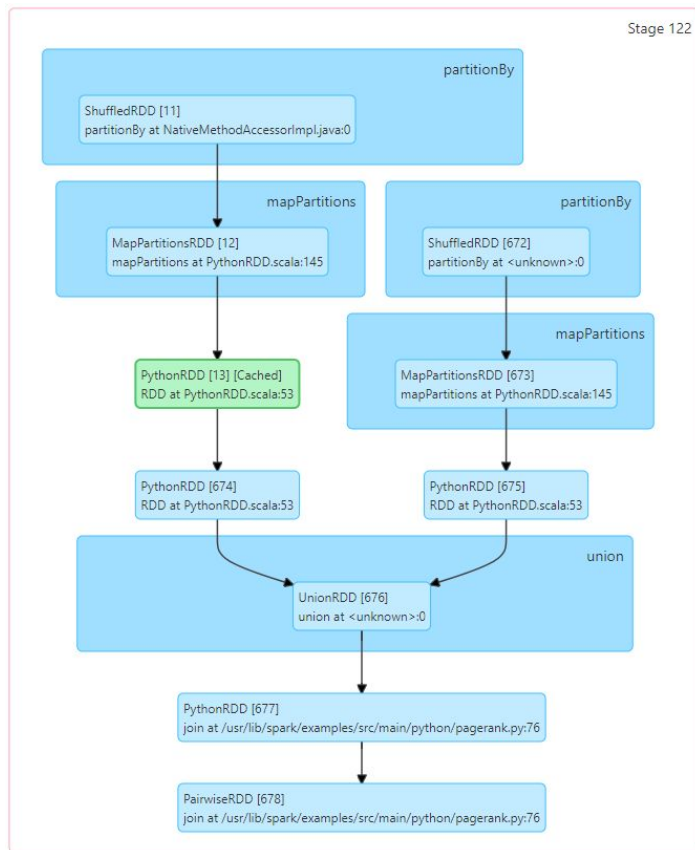
## Vantagens

- Trata a estrutura de dados tanto como grafos comuns, quanto como conjuntos separados de arestas e vértices que podem ter operações feitas de forma paralela (*map*, *join*, *transform*).
- Lida bem com *property graphs*.

## Desvantagens

- Latência alta
- Alto consumo de memória
- Não é eficiente quando há um grande número de iterações, pois pode haver sobrecarga de RDDs ([Fonte](#)).

# DAG (DIRECTED ACYCLIC GRAPH)





# Onde é usado

- Netflix - recomendação (junto com machine learning)
- <https://pt.slideshare.net/SessionsEvents/ehtsham-elahi-senior-research-engineer-personalization-science-and-engineering-group-at-netflix-at-mlconf-sea-50115>
-

# A API





[org.apache.spark.graphx](http://org.apache.spark.graphx)

## Graph

Related Docs: [object Graph](#) | [package graphx](#)

`abstract class Graph[VD, ED] extends Serializable`

The Graph abstractly represents a graph with arbitrary objects associated with vertices and edges. The graph provides basic operations to access and manipulate the data associated with vertices and edges as well as the underlying structure. Like Spark RDDs, the graph is a functional data-structure in which mutating operations return new graphs.

**VD**      the vertex attribute type

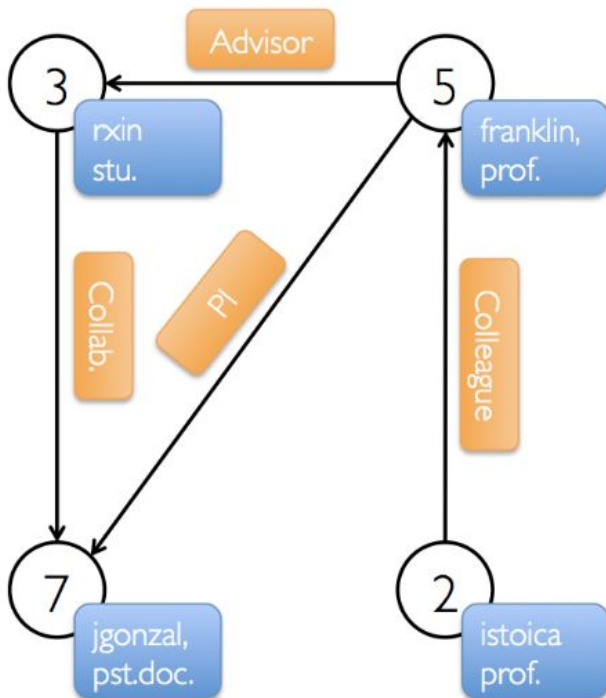
**ED**      the edge attribute type

- Property graph: multi grafo direcionado com objetos anexados a cada vértice e aresta.
- São imutáveis, distribuídos e tolerante a falhas.
- Cada partição do grafo pode ser recriada em uma máquina diferente caso haja falha.

```
class Graph[VD, ED] {  
    val vertices: VertexRDD[VD]  
    var edges: EdgeRDD[ED]  
}
```

# Graphx - exemplificando

## Property Graph



## Vertex Table

| Id | Property (V)          |
|----|-----------------------|
| 3  | (rxin, student)       |
| 7  | (jgonzal, postdoc)    |
| 5  | (franklin, professor) |
| 2  | (istoica, professor)  |

## Edge Table

| SrcId | DstId | Property (E) |
|-------|-------|--------------|
| 3     | 7     | Collaborator |
| 5     | 3     | Advisor      |
| 2     | 5     | Colleague    |
| 5     | 7     | PI           |

## Vertex Table

| Id | Property (V)          |
|----|-----------------------|
| 3  | (rxin, student)       |
| 7  | (jgonzal, postdoc)    |
| 5  | (franklin, professor) |
| 2  | (istoica, professor)  |

```
// Create an RDD for the vertices
```

```
val users: RDD[(VertexId, (String, String))] =  
  sc.parallelize(Array((3L, ("rxin", "student")), (7L, ("jgonzal", "postdoc")),  
    (5L, ("franklin", "prof")), (2L, ("istoica", "prof"))))
```

# Graphx - exemplificando

Edge Table

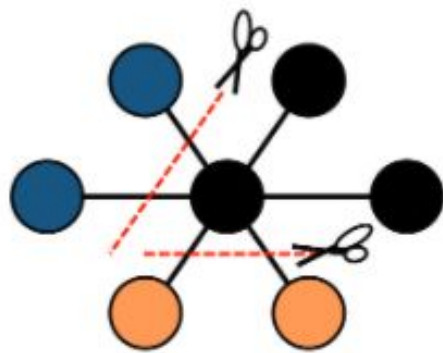
| SrcId | DstId | Property (E) |
|-------|-------|--------------|
| 3     | 7     | Collaborator |
| 5     | 3     | Advisor      |
| 2     | 5     | Colleague    |
| 5     | 7     | PI           |

```
// Create an RDD for edges
```

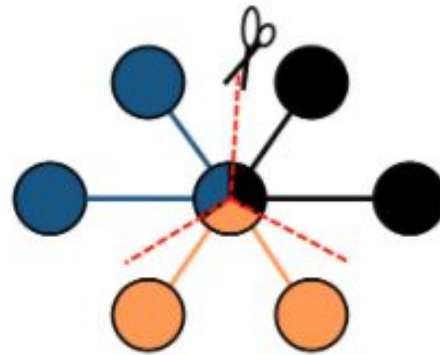
```
val relationships: RDD[Edge[String]] =
```

```
    sc.parallelize(Array(Edge(3L, 7L, "collab"),    Edge(5L, 3L, "advisor"),  
                        Edge(2L, 5L, "colleague"), Edge(5L, 7L, "pi")))
```

# PROPERTY GRAPHS - otimizações



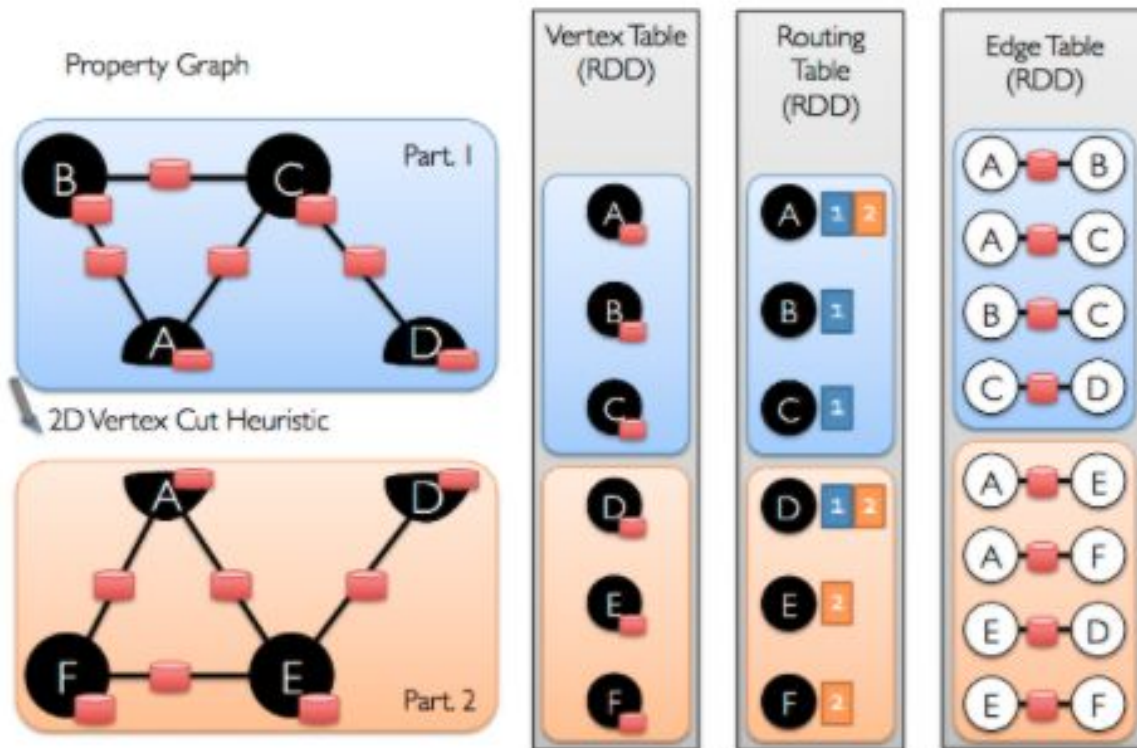
Edge Cut



Vertex Cut



# PROPERTY GRAPHS - otimizações



# PROPERTY GRAPHS - otimizações



org.apache.spark.graphx

## PartitionStrategy

Companion [trait PartitionStrategy](#)

object **PartitionStrategy** extends Serializable

Collection of built-in [PartitionStrategy](#) implementations.

**Source** [PartitionStrategy.scala](#)

Linear Supertypes

Filter all members

### Value Members

def **fromString**(s: String): [PartitionStrategy](#)

Returns the PartitionStrategy with the specified name.

object [CanonicalRandomVertexCut](#) extends [PartitionStrategy](#) with Product with Serializable

Assigns edges to partitions by hashing the source and destination vertex IDs in a canonical direction, resulting in a random vertex cut that colocates all edges between two vertices, regardless of direction.

object [EdgePartition1D](#) extends [PartitionStrategy](#) with Product with Serializable

Assigns edges to partitions using only the source vertex ID, colocating edges with the same source.

object [EdgePartition2D](#) extends [PartitionStrategy](#) with Product with Serializable

Assigns edges to partitions using a 2D partitioning of the sparse edge adjacency matrix, guaranteeing a  $2 * \sqrt{\text{numParts}}$  bound on vertex replication.

object [RandomVertexCut](#) extends [PartitionStrategy](#) with Product with Serializable

Assigns edges to partitions by hashing the source and destination vertex IDs, resulting in a random vertex cut that colocates all same-direction edges between two vertices.



# FRAMEWORKS RELACIONADOS



- Sistema da Google, alternativo ao Spark GraphX, para processamento de grafos de larga escala;
- Escalável e tolerante a falhas;
- Alimenta o PageRank da ferramenta de pesquisa;
- Inspiração para o Apache Giraph, que o Facebook usa para analisar seu grafo de rede social;

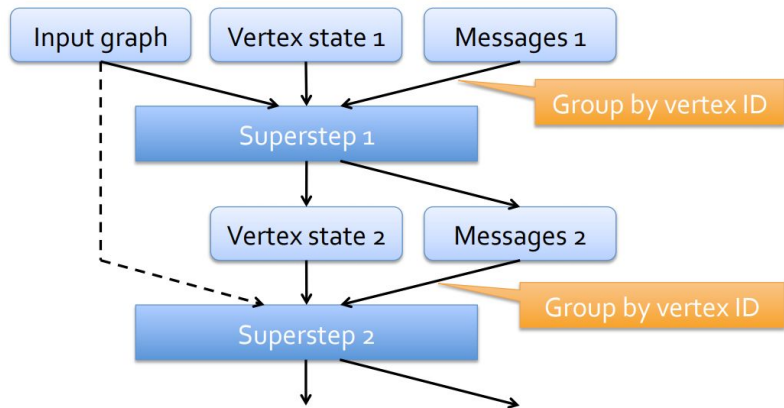
- Usado no Facebook para analisar o grafo de rede social formado pelos usuários e suas conexões;
- Giraph se originou como a contraparte open-source para o Pregel;
- Adiciona vários recursos além do modelo Pregel básico, incluindo computação mestre, agregadores fragmentados, entrada de conjunto de arestas, computação fora do núcleo;

# Pregel & GraphX- comparação

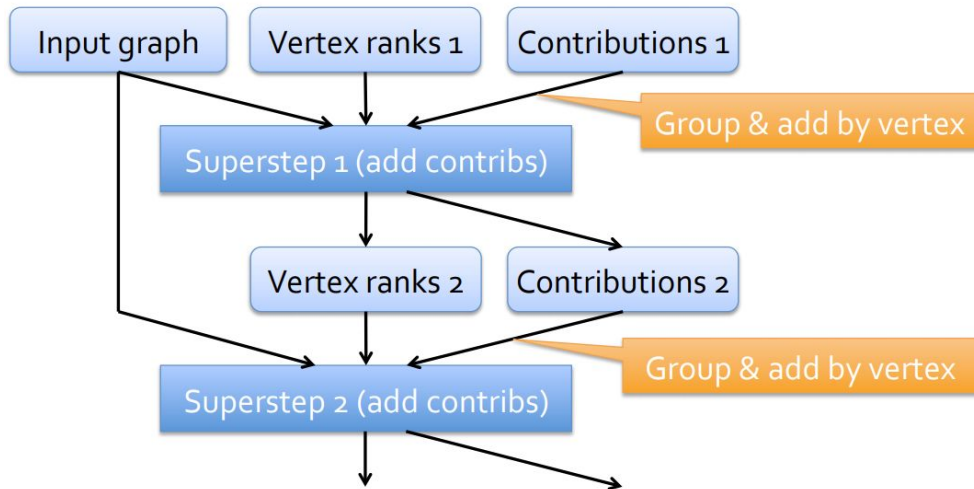
- Expõem API específicas para simplificar programação em grafos
- Se baseiam no padrão em que a computação depende apenas dos vizinhos
- Expressam os jobs como grafos de operadores de alto nível
  - O sistema escolhe como dividir cada operador em tarefas e onde executar cada tarefa
  - Executam partes para recuperação de falhas duas vezes
- GraphX roda o modelo do Pregel por trás de suas tabelas de vértices e arestas

# Pregel & GraphX- comparação

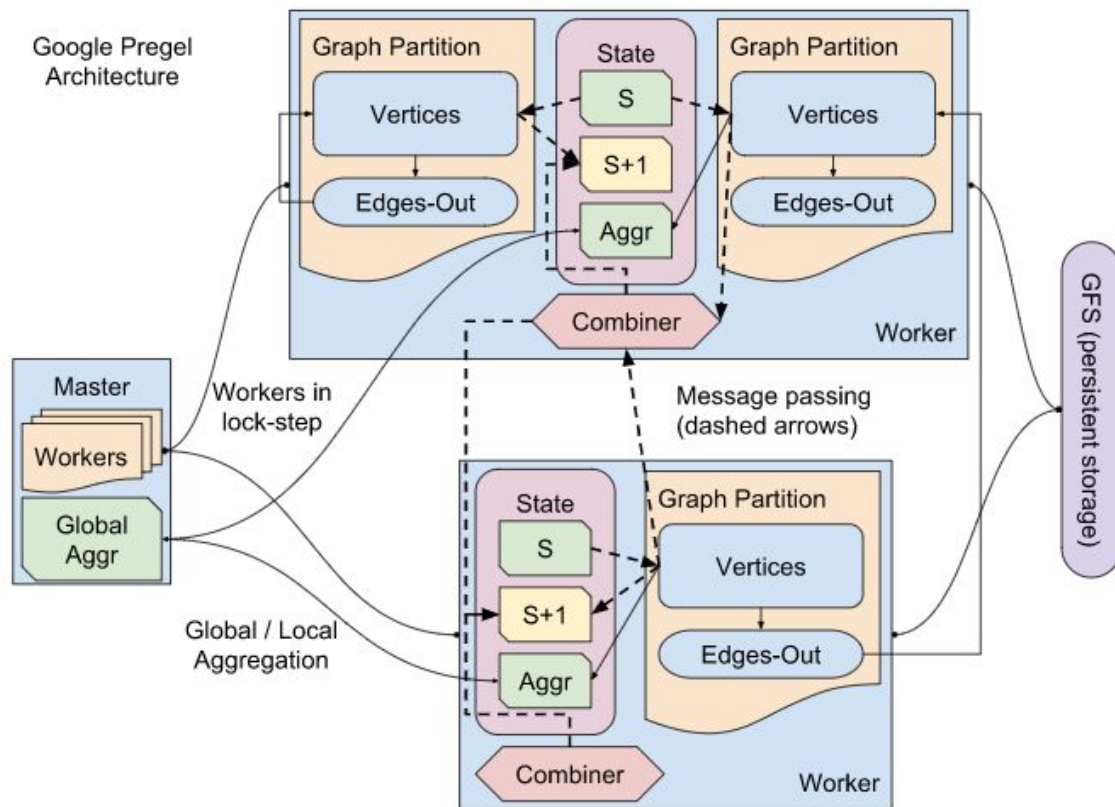
## Pregel Data Flow



## PageRank in Pregel

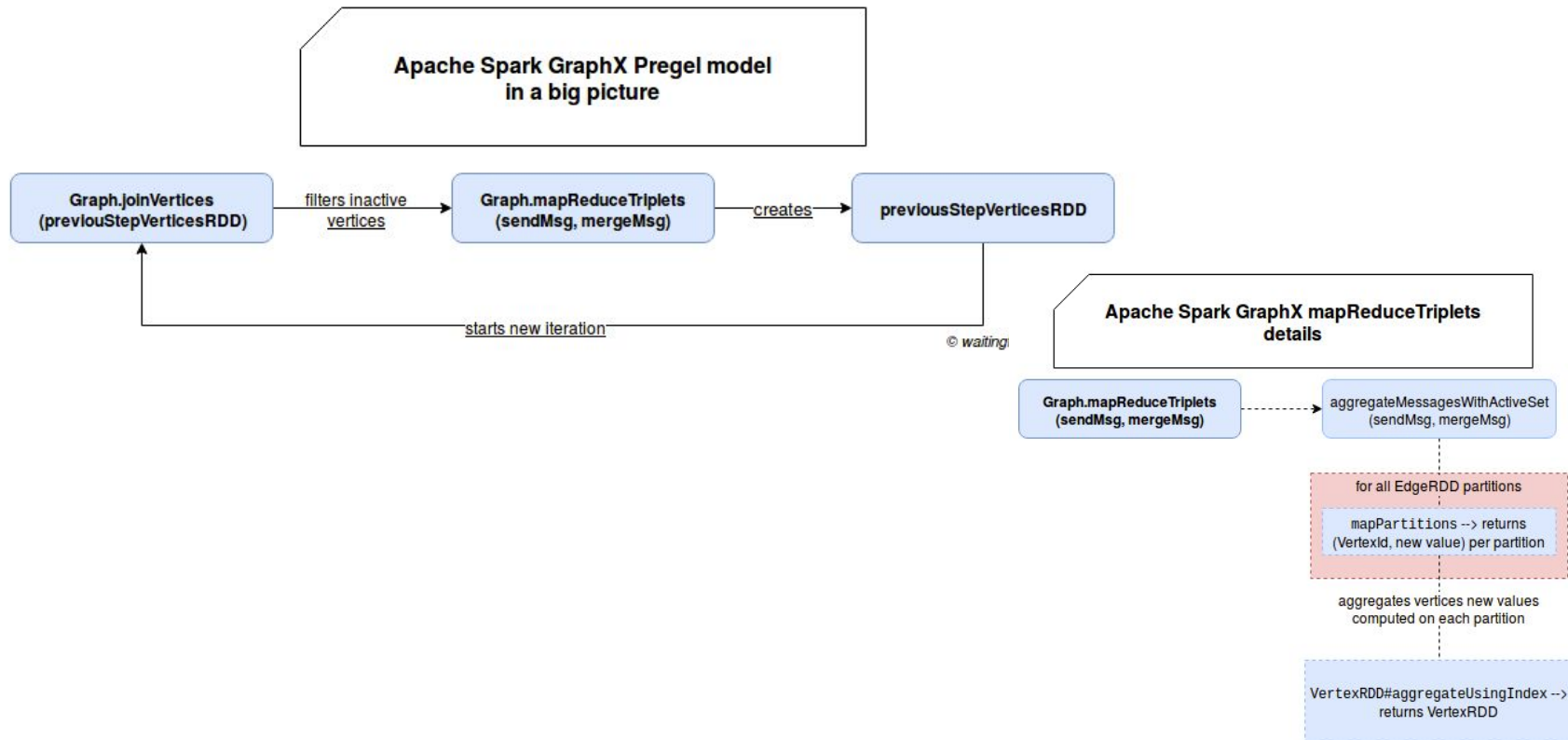


# Pregel & GraphX- comparação





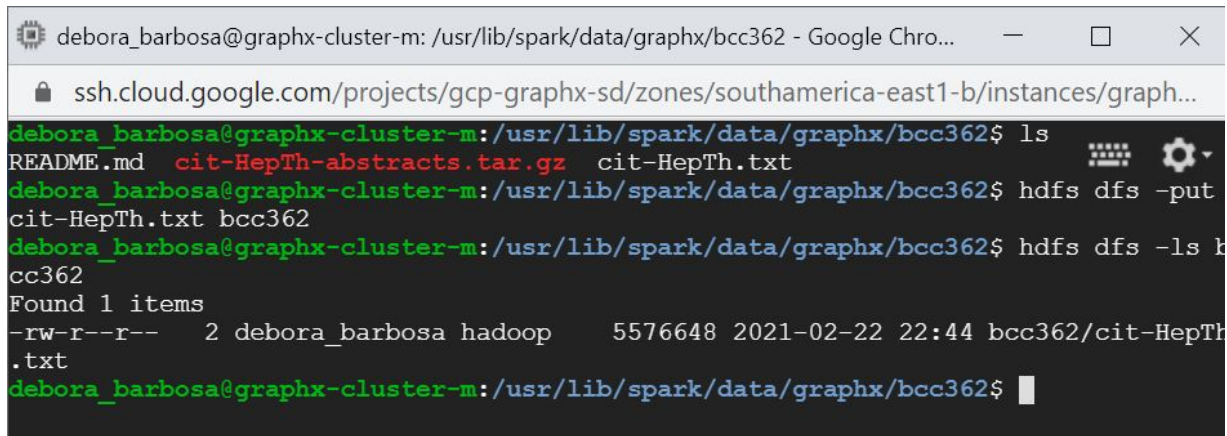
# Pregel & GraphX- comparação



# FUNCIONAMENTO



# Funcionamento



The screenshot shows a terminal window with a title bar indicating the user is 'debora\_barbosa' on a 'graphx-cluster-m' machine, with the path '/usr/lib/spark/data/graphx/bcc362'. The browser address bar shows an SSH connection to a Google Cloud project. The terminal content shows the user listing files, uploading a file to HDFS, and then listing the contents of the HDFS directory.

```
debora_barbosa@graphx-cluster-m: /usr/lib/spark/data/graphx/bcc362$ ls
README.md  cit-HepTh-abstracts.tar.gz  cit-HepTh.txt
debora_barbosa@graphx-cluster-m: /usr/lib/spark/data/graphx/bcc362$ hdfs dfs -put
cit-HepTh.txt bcc362
debora_barbosa@graphx-cluster-m: /usr/lib/spark/data/graphx/bcc362$ hdfs dfs -ls h
cc362
Found 1 items
-rw-r--r--  2 debora_barbosa  hadoop    5576648  2021-02-22  22:44  bcc362/cit-HepTh
.txt
debora_barbosa@graphx-cluster-m: /usr/lib/spark/data/graphx/bcc362$
```

```
# Directed graph (each unordered pair of nodes is saved once): Cit-HepTh.txt
```

```
# Paper citation network of Arxiv High Energy Physics Theory category
```

```
# Nodes: 27770 Edges: 352807
```

```
# FromNodeId ToNodeId
```

```
1001 9304045
```

```
1001 9308122
```

```
1001 9309097
```

```
1001 9311042
```

```
1001 9401139
```

```
1001 9404151
```

```
1001 9407087
```

```
1001 9408099
```

```
1001 9501030
```

```
1001 9503124
```

```
1001 9504090
```

```
1001 9504145
```

```
1001 9505025
```

```
1001 9505054
```

```
1001 9505105
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```
1001 9505162
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1001 9506048
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1001 9506112
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1001 9506144
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```
1001 9507050
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```
1001 9507158
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```
1001 9508094
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```
1001 9508155
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```
1001 9510142
```

```
1001 9510225
```

```
1001 9510234
```

```
1001 9511030
```

```
1001 9511171
```

```
1001 9601108
```

```
1001 9602022
```

```
1001 9602114
```

```
1001 9603003
```

```
^G Get Help
```

```
^O Write Out
```

```
^W Where Is
```

```
^K Cut Text
```

```
^J Justify
```

```
^C Cur Pos
```

```
M-U Undo
```

```
^X Exit
```

```
^R Read File
```

```
^_ Replace
```

```
^U Uncut Text
```

```
^T To Spell
```

```
^_ Go To Line
```

```
M-E Redo
```

- ☐ Off Heap Memory
- ☐ Peak JVM Memory OnHeap / OffHeap
- ☐ Peak Execution Memory OnHeap / OffHeap
- ☐ Peak Storage Memory OnHeap / OffHeap
- ☐ Peak Pool Memory Direct / Mapped
- ☐ Resources
- ☐ Resource Profile Id

## Summary

|                  | RDD Blocks | Storage Memory   | Disk Used | Cores | Active Tasks | Failed Tasks | Complete Tasks | Total Tasks | Task Time (GC Time) | Input     | Shuffle Read | Shuffle Write | Excluded |
|------------------|------------|------------------|-----------|-------|--------------|--------------|----------------|-------------|---------------------|-----------|--------------|---------------|----------|
| <b>Active(9)</b> | 0          | 0.0 B / 11.6 GiB | 0.0 B     | 8     | 1            | 0            | 7936           | 7937        | 20 min (1.4 min)    | 130.7 MiB | 360.3 MiB    | 363 MiB       | 0        |
| <b>Dead(1)</b>   | 0          | 0.0 B / 1.2 GiB  | 0.0 B     | 1     | 0            | 0            | 1              | 1           | 6 s (0.2 s)         | 675 KiB   | 0.0 B        | 352.3 KiB     | 0        |
| <b>Total(10)</b> | 0          | 0.0 B / 12.8 GiB | 0.0 B     | 9     | 1            | 0            | 7937           | 7938        | 20 min (1.4 min)    | 131.4 MiB | 360.3 MiB    | 363.4 MiB     | 0        |

## Executors

Show  entries

Search:

| Executor ID | Address   | Status | RDD Blocks | Storage Memory  | Disk Used | Cores | Active Tasks | Failed Tasks | Complete Tasks | Total Tasks | Task Time (GC Time) | Input     | Shuffle Read | Shuffle Write | Logs   |
|-------------|---|--------|------------|-----------------|-----------|-------|--------------|--------------|----------------|-------------|---------------------|-----------|--------------|---------------|--|
| driver      | graphx-cluster-m.c.gcp-graphx-sd.internal:37727   | Active | 0          | 0.0 B / 1.8 GiB | 0.0 B     | 0     | 0            | 0            | 0              | 0           | 0.0 ms (0.0 ms)     | 0.0 B     | 0.0 B        | 0.0 B         |  |
| 1           | graphx-cluster-w-0.c.gcp-graphx-sd.internal:43175 | Dead   | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 0            | 0            | 1              | 1           | 6 s (0.2 s)         | 675 KiB   | 0.0 B        | 352.3 KiB     | <a href="#">stdout</a><br><a href="#">stderr</a> |
| 2           | graphx-cluster-w-2.c.gcp-graphx-sd.internal:40085 | Active | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 1            | 0            | 4231           | 4232        | 9.5 min (40 s)      | 130.7 MiB | 215.9 MiB    | 265.4 MiB     | <a href="#">stdout</a><br><a href="#">stderr</a> |
| 3           | graphx-cluster-w-1.c.gcp-graphx-sd.internal:46527 | Active | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 0            | 0            | 79             | 79          | 24 s (0.3 s)        | 0.0 B     | 2.8 MiB      | 2.4 MiB       | <a href="#">stdout</a><br><a href="#">stderr</a> |
| 4           | graphx-cluster-w-0.c.gcp-graphx-sd.internal:34099 | Active | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 0            | 0            | 82             | 82          | 26 s (0.5 s)        | 0.0 B     | 3 MiB        | 2.5 MiB       | <a href="#">stdout</a><br><a href="#">stderr</a> |
| 5           | graphx-cluster-w-3.c.gcp-graphx-sd.internal:44139 | Active | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 0            | 0            | 80             | 80          | 30 s (0.5 s)        | 0.0 B     | 2.9 MiB      | 2.5 MiB       | <a href="#">stdout</a><br><a href="#">stderr</a> |
| 6           | graphx-cluster-w-3.c.gcp-graphx-sd.internal:37297 | Active | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 0            | 0            | 79             | 79          | 28 s (0.6 s)        | 0.0 B     | 2.9 MiB      | 2.3 MiB       | <a href="#">stdout</a><br><a href="#">stderr</a> |
| 7           | graphx-cluster-w-1.c.gcp-graphx-sd.internal:46207 | Active | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 0            | 0            | 83             | 83          | 26 s (0.3 s)        | 0.0 B     | 3.1 MiB      | 2.5 MiB       | <a href="#">stdout</a><br><a href="#">stderr</a> |
| 8           | graphx-cluster-w-2.c.gcp-graphx-sd.internal:42305 | Active | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 0            | 0            | 3220           | 3220        | 7.7 min (41 s)      | 0.0 B     | 126.7 MiB    | 83.1 MiB      | <a href="#">stdout</a><br><a href="#">stderr</a> |
| 9           | graphx-cluster-w-0.c.gcp-graphx-sd.internal:33975 | Active | 0          | 0.0 B / 1.2 GiB | 0.0 B     | 1     | 0            | 0            | 82             | 82          | 24 s (0.5 s)        | 0.0 B     | 2.9 MiB      | 2.4 MiB       | <a href="#">stdout</a><br><a href="#">stderr</a> |

Showing 1 to 10 of 10 entries

Previous **1** Next

Não seguro | 34.95.148.125:18080/history/application\_1614206849165\_0003/stages/

Spark3.1.1

Jobs

Stages

Storage

Environment

Executors

PythonPageRank application UI

## Stages for All Jobs

Completed Stages: 124  
Skipped Stages: 78  
Failed Stages: 1

Completed Stages (124)

Page: 1 2 > 2 Pages. Jump to 1 . Show 100 items in a page. Go

| Stage Id | Description   | Submitted                    | Duration | Tasks: Succeeded/Total | Input | Output | Shuffle Read | Shuffle Write |
|----------|---|------------------------------|----------|------------------------|-------|--------|--------------|---------------|
| 123      | reduceByKey at /usr/lib/spark/examples/src/main/python/pagerank.py:80 | +details 2021/02/24 23:37:24 | 12 s     | 124/124                |       |        | 3.1 MiB      | 4.7 MiB       |



3.1.1

Jobs

Stages

Storage

Environment

Executors

PythonPageRank application UI

## Details for Stage 123 (Attempt 0)

Resource Profile Id: 0

Total Time Across All Tasks: 26 s

Locality Level Summary: Node local: 112; Rack local: 12

Shuffle Read Size / Records: 3.1 MiB / 7441

Shuffle Write Size / Records: 4.7 MiB / 30543

Associated Job Ids: 0

DAG Visualization

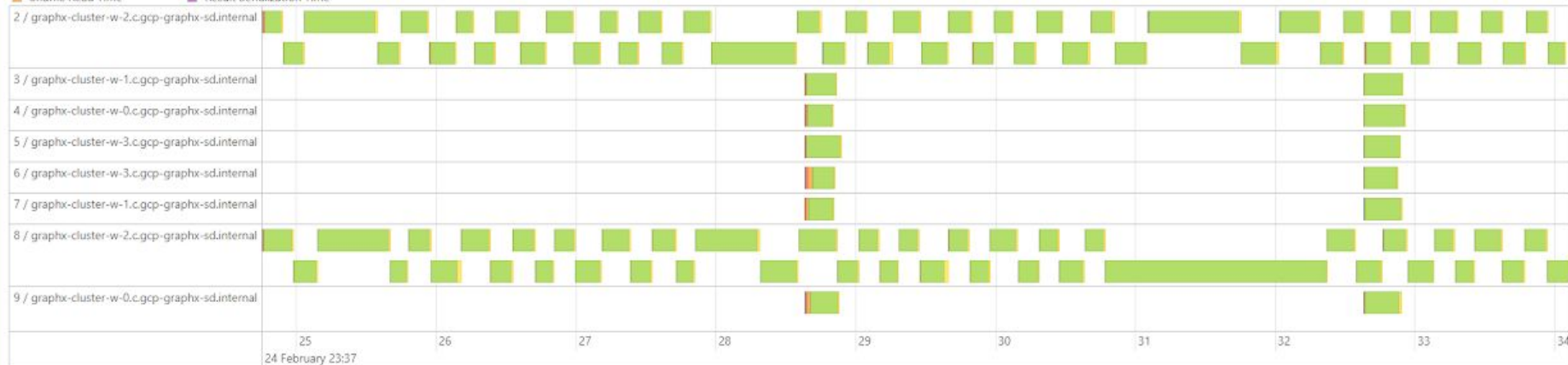
Show Additional Metrics

Event Timeline

☐ Enable zooming

■ Scheduler Delay
 ■ Executor Computing Time
 ■ Getting Result Time
 ■ Task Deserialization Time
 ■ Shuffle Write Time
 ■ Result Serialization Time
 ■ Shuffle Read Time

Tasks: 124. 2 Pages. Jump to  . Show  items in a page. Go



### Summary Metrics for 124 Completed Tasks

| Metric   | Min    | 25th percentile | Median | 75th percentile | Max |
|----------|--------|-----------------|--------|-----------------|-----|
| Duration | 0.1 s  | 0.1 s           | 0.2 s  | 0.2 s           | 2 s |
| GC Time  | 0.0 ms | 0.0 ms          | 0.0 ms | 0.0 ms          | 1 s |

— — —

| número de iterações | tempo de execução |
|---------------------|-------------------|
| 1                   | 26s               |
| 10                  | 50s               |
| 50                  | 6.7min            |
| 100                 | 10min             |



# Referências

Michael Malak, Robin East - Spark GraphX in Action (2016, Manning Publications) - libgen.lc

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Ankur Dave: ankur is a third-year phd student advised by ion stoica in the UC Berkeley AMPLab. He's a Spark committer and a maintainer for GraphX

<https://spark.apache.org/docs/latest/graphx-programming-guide.html#vertexrdds> - graphx programming guide

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