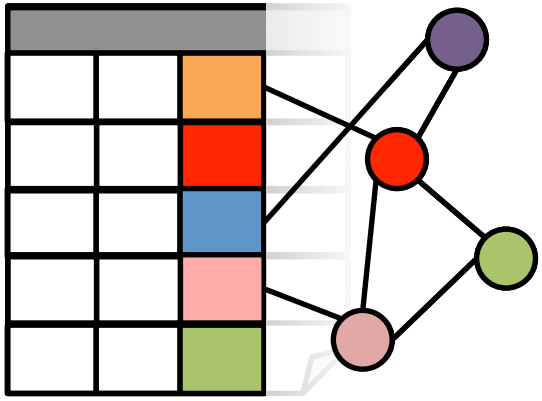
GraphX



Graph Analytics in Spark

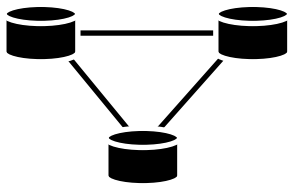
Ankur Dave
Graduate Student, UC Berkeley AMPLab

Joint work with Joseph Gonzalez, Reynold Xin, Daniel Crankshaw, Michael Franklin, and Ion Stoica

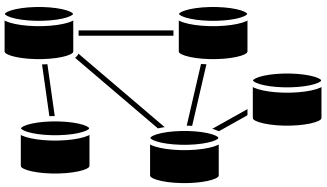
amplab
UC BERKELEY

Machine Learning Landscape

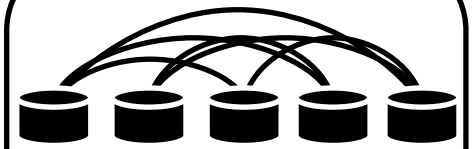
Model & Dependencies



Small & Dense

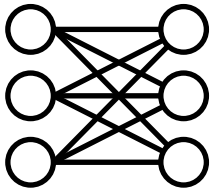


Sparse

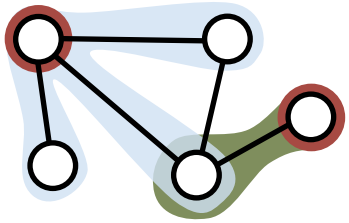


Large & Dense

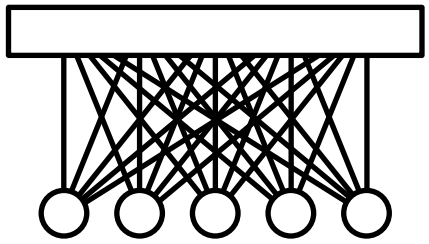
Architecture



MapReduce



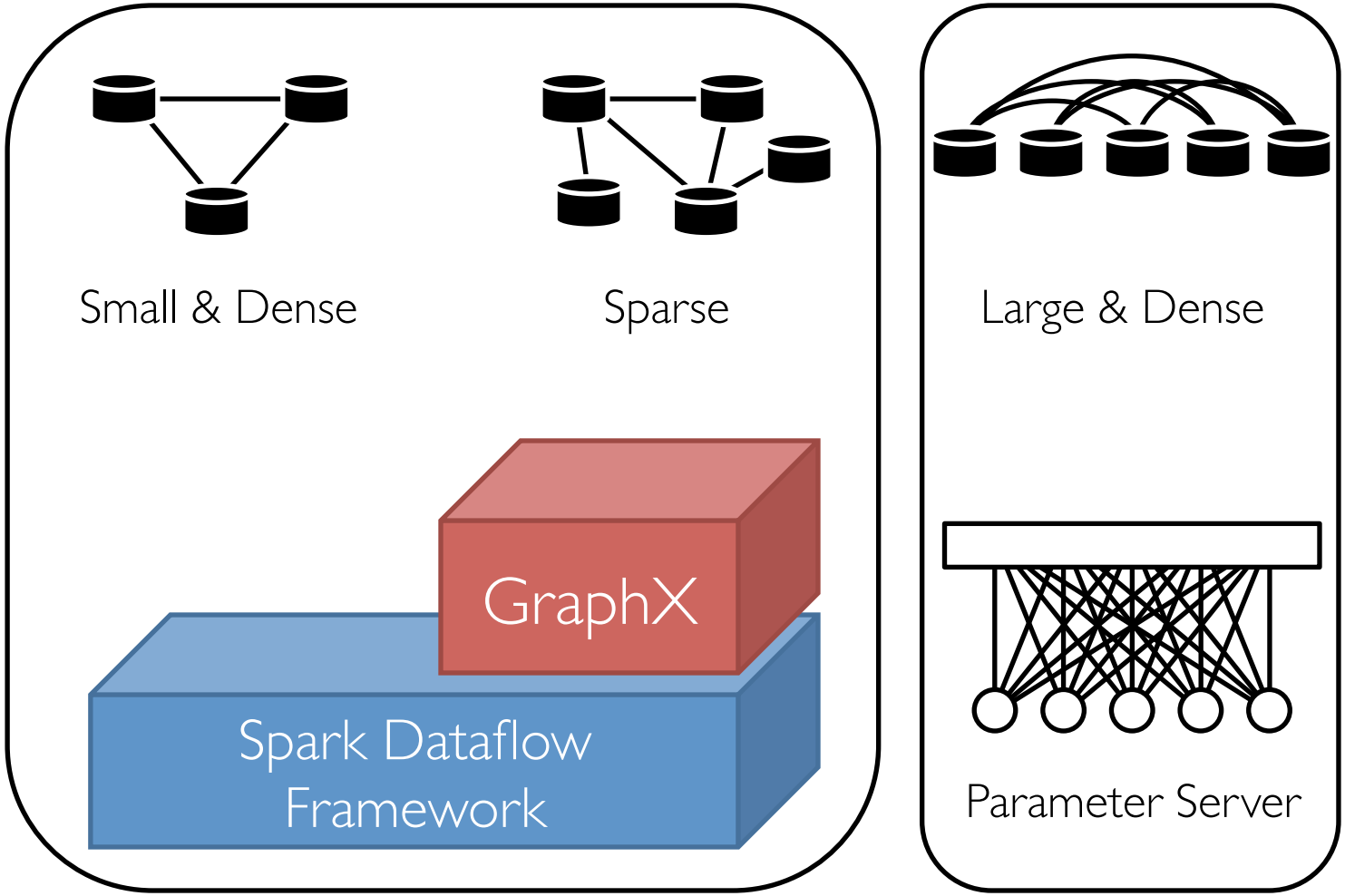
Graph-Parallel



Parameter Server

Machine Learning Landscape

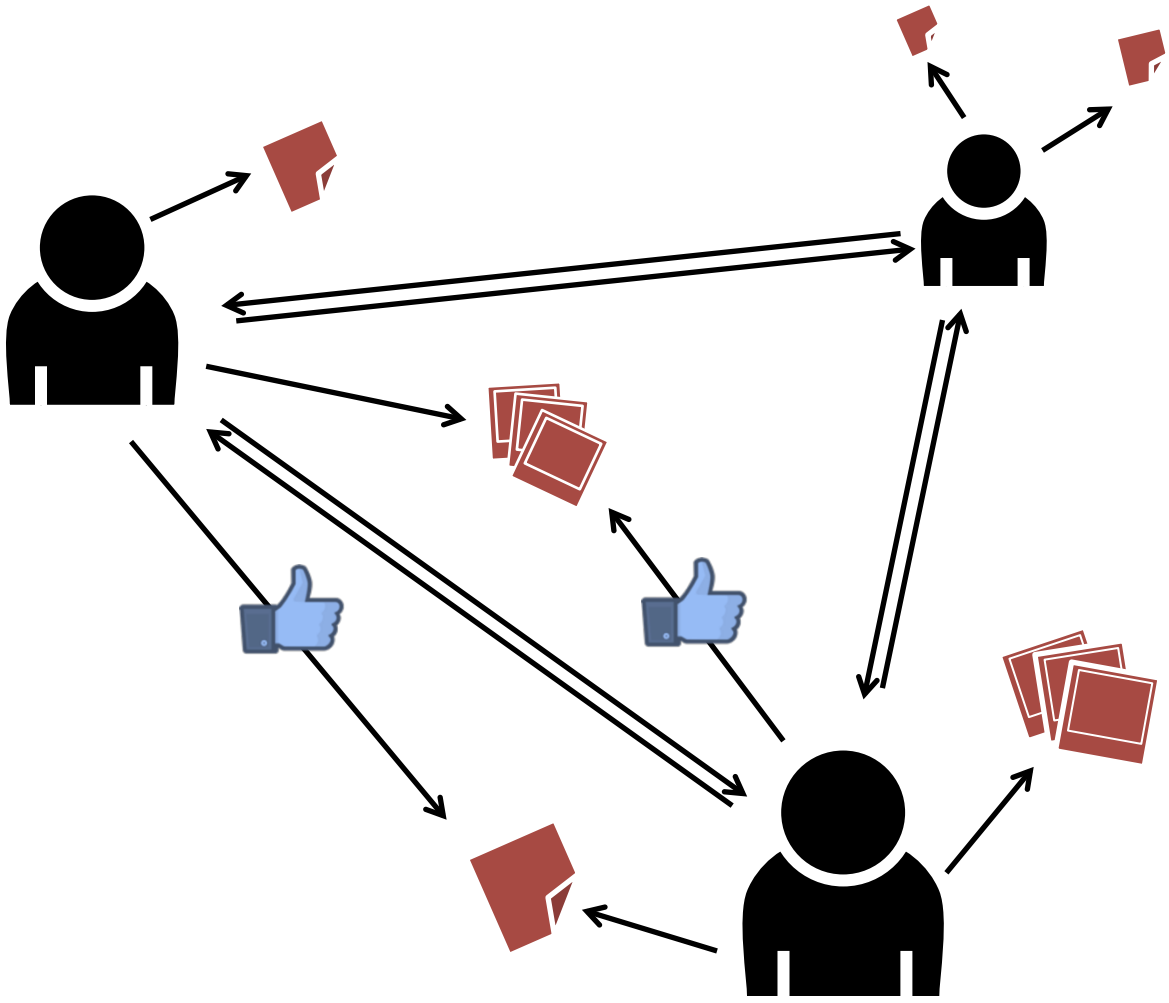
Model & Dependencies



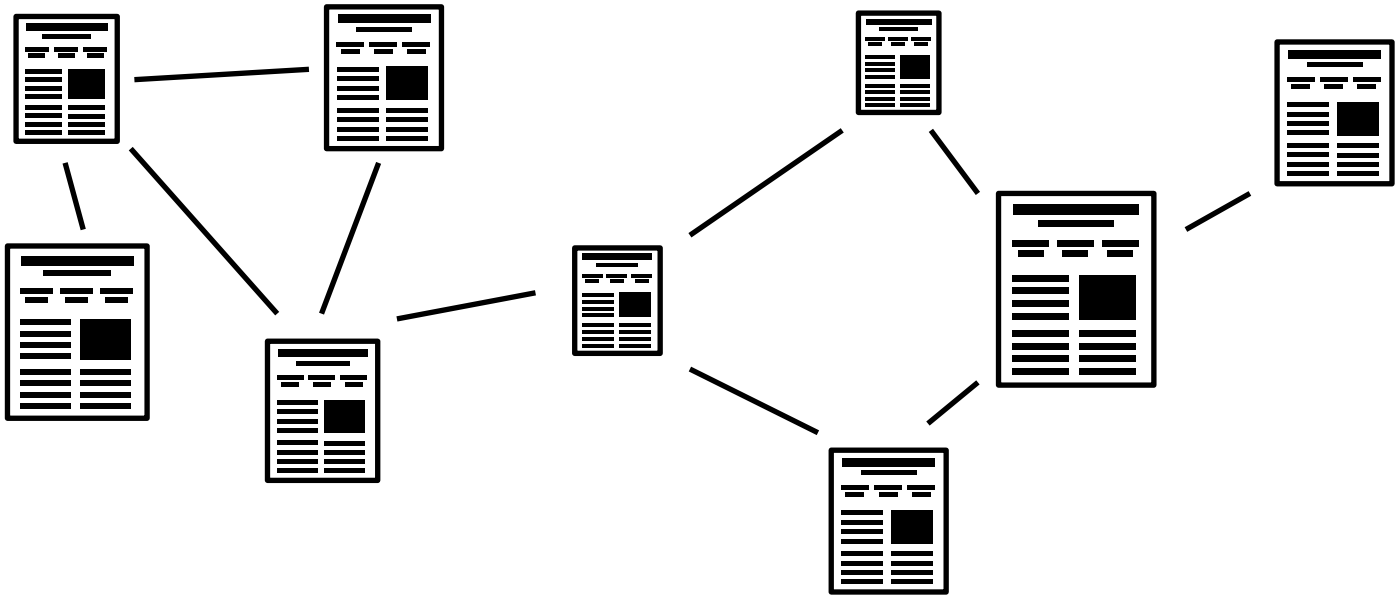
Architecture

Graphs

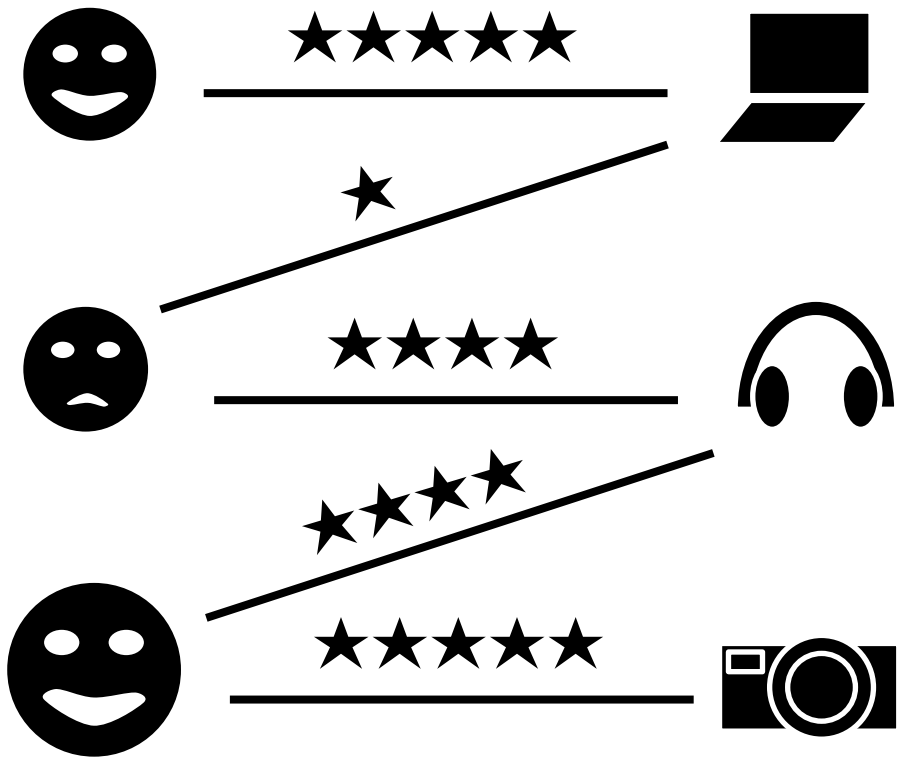
Social Networks



Web Graphs

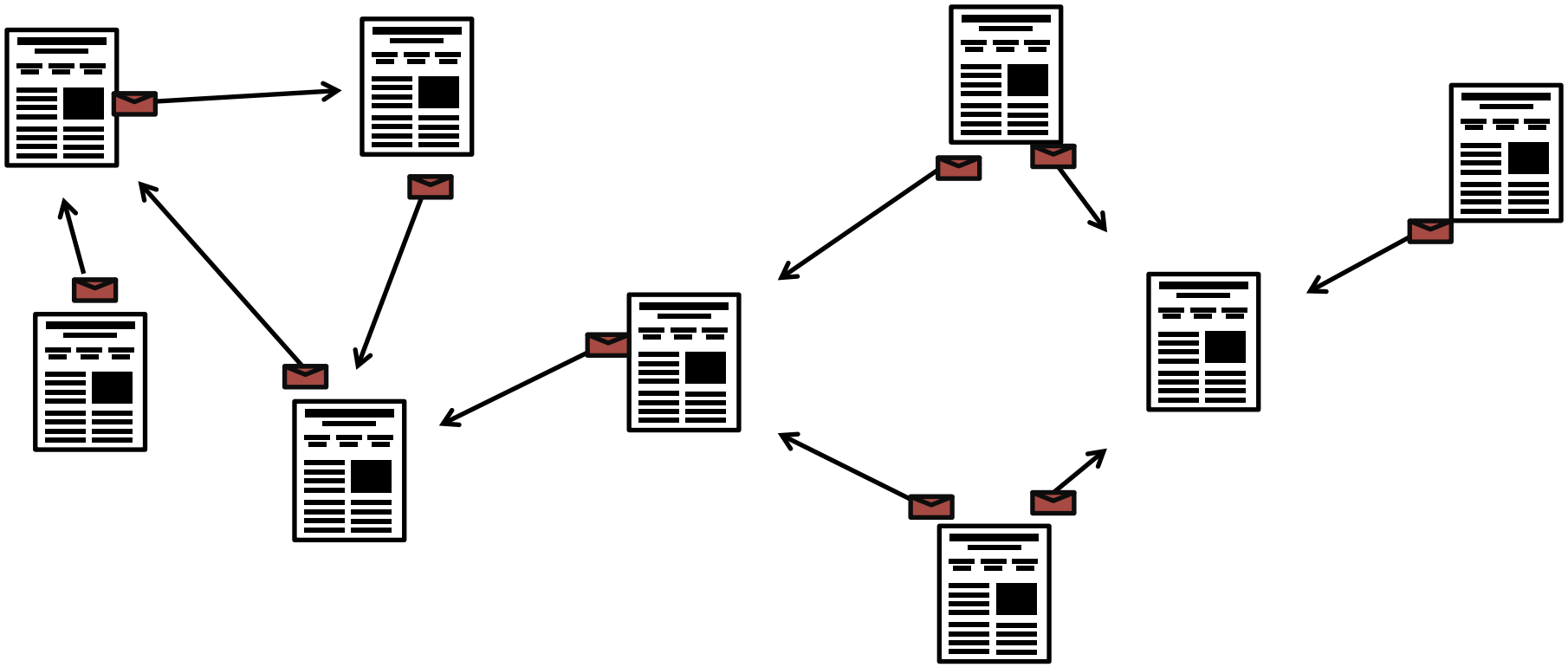


User-Item Graphs

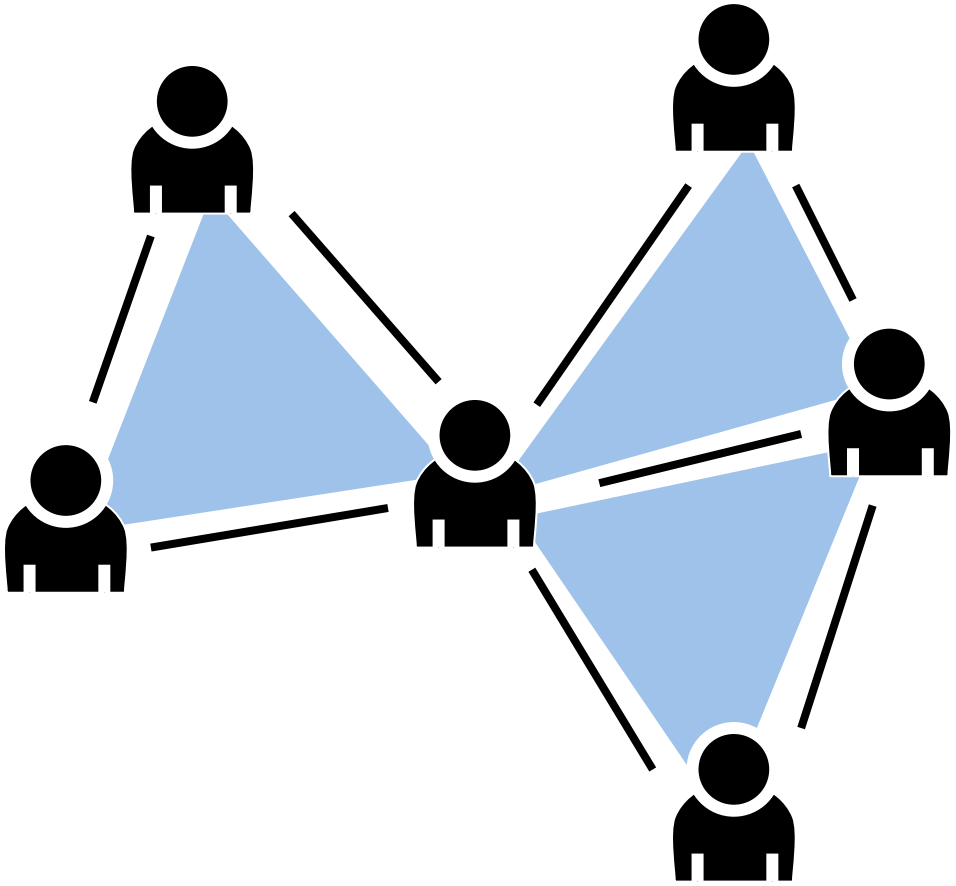


Graph Algorithms

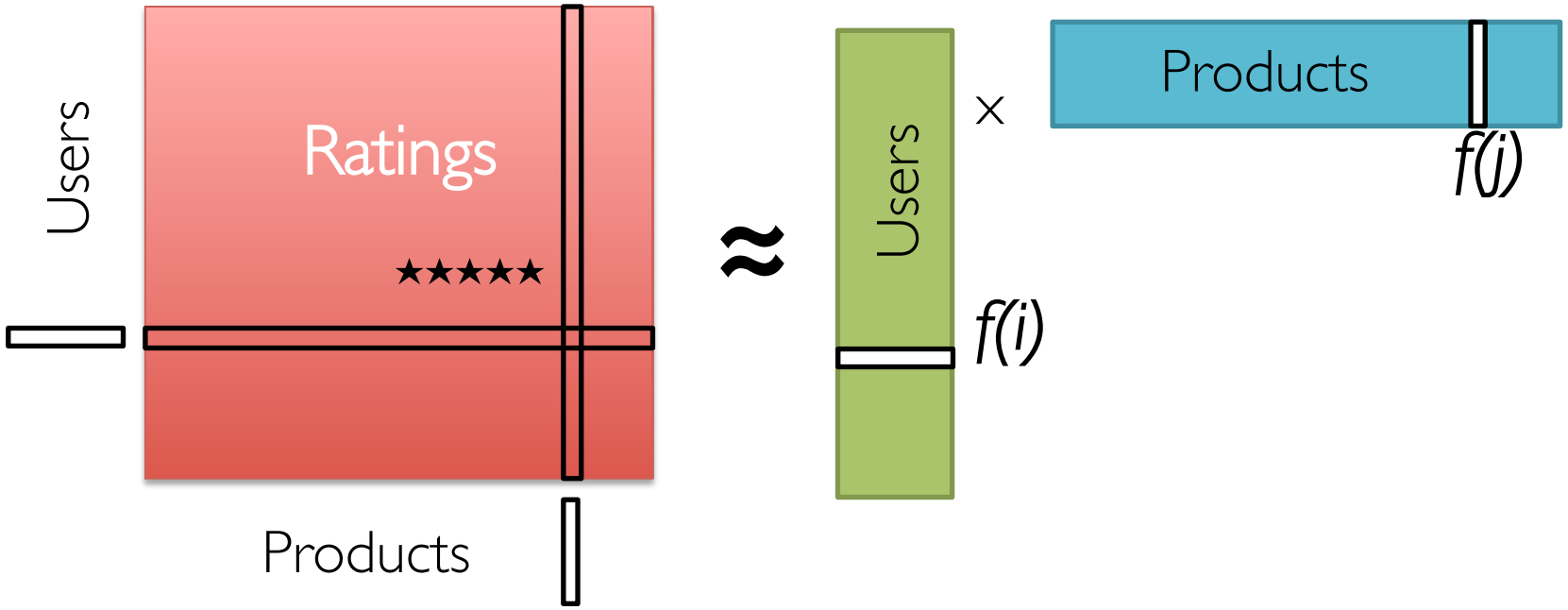
PageRank



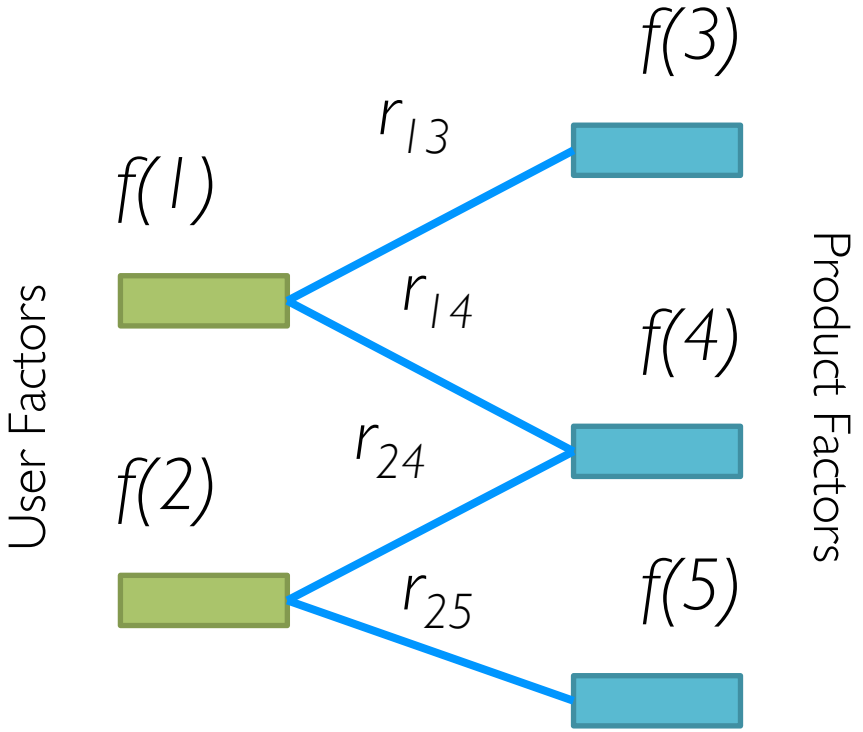
Triangle Counting



Collaborative Filtering

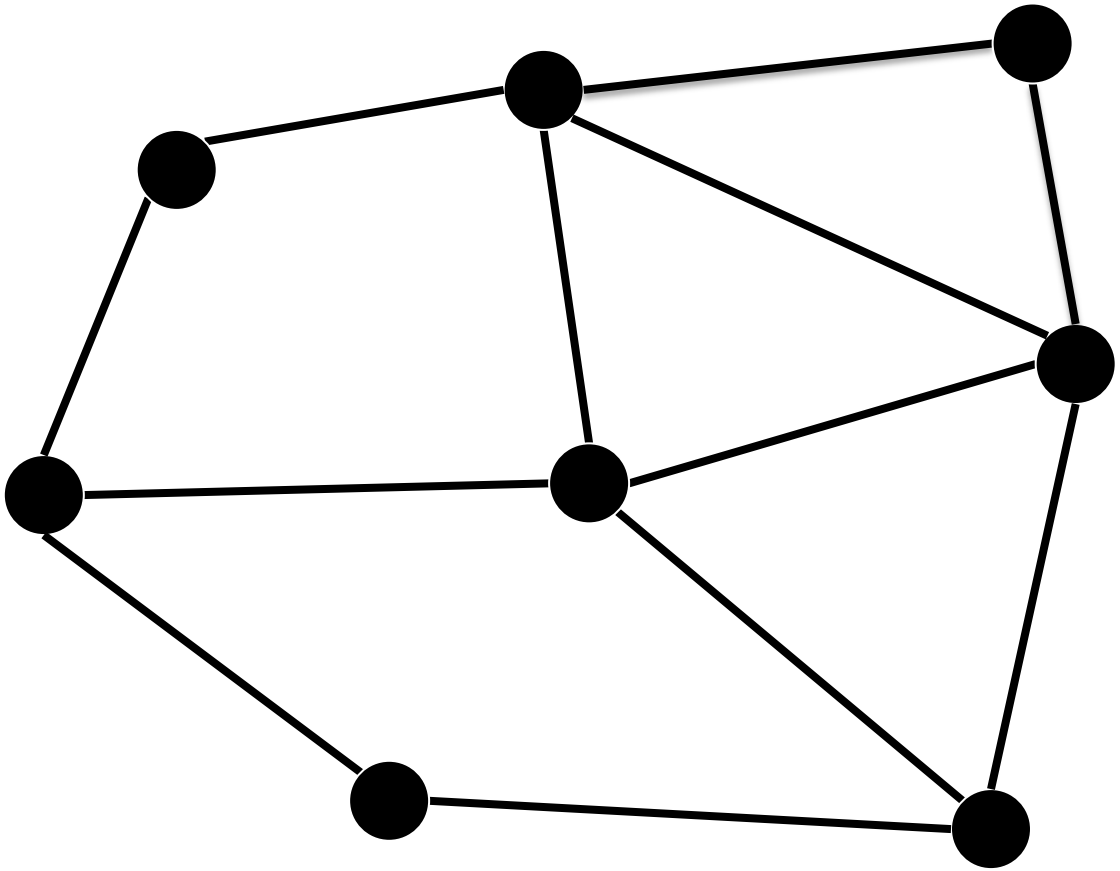


Collaborative Filtering

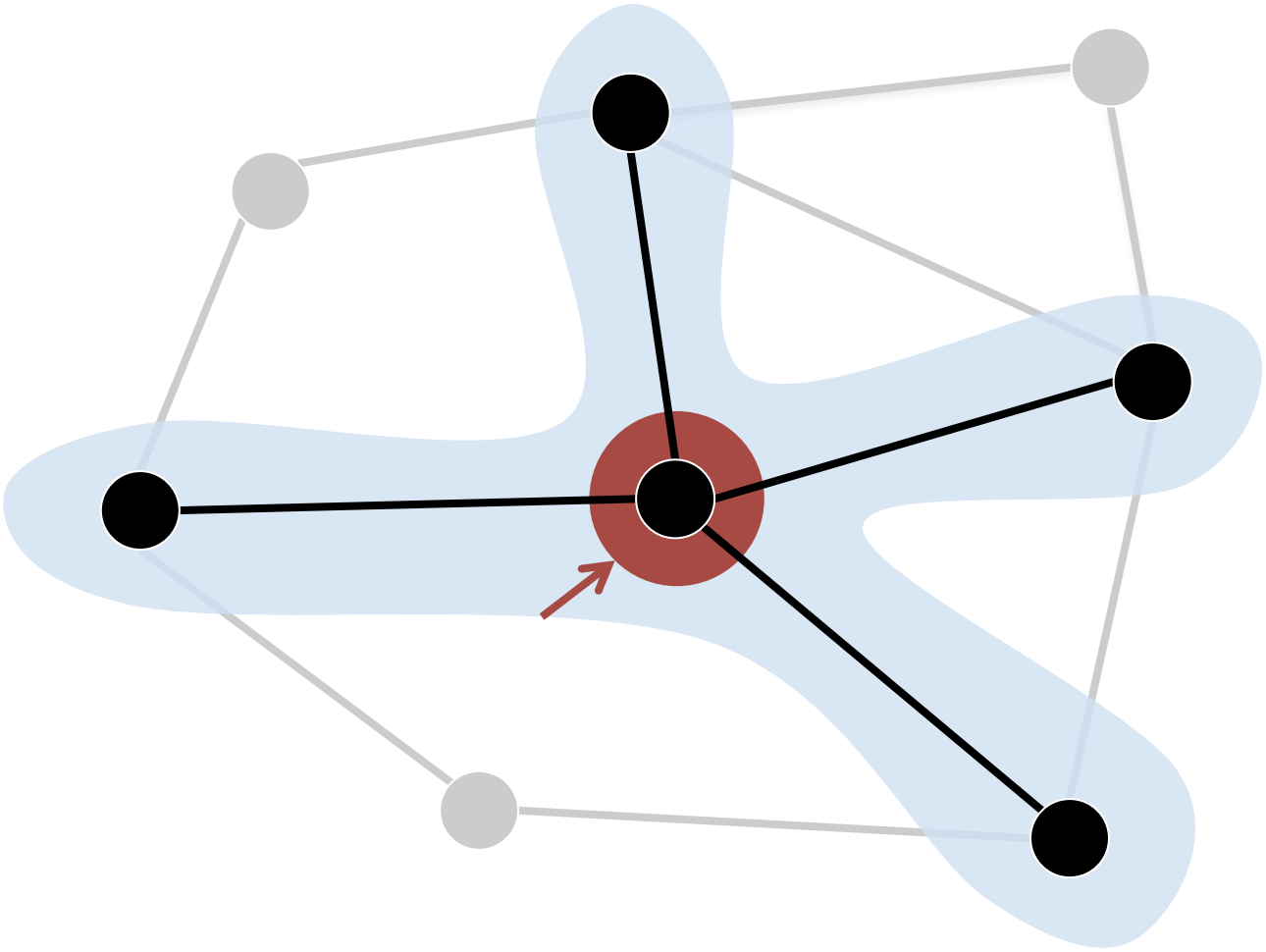


$$f[i] = \arg\min_{w \in \mathbb{R}^d} \sum_{j \in \mathrm{Nbrs}(i)} \left(r_{ij} - w^T f[j]\right)^2 + \lambda ||w||_2^2$$

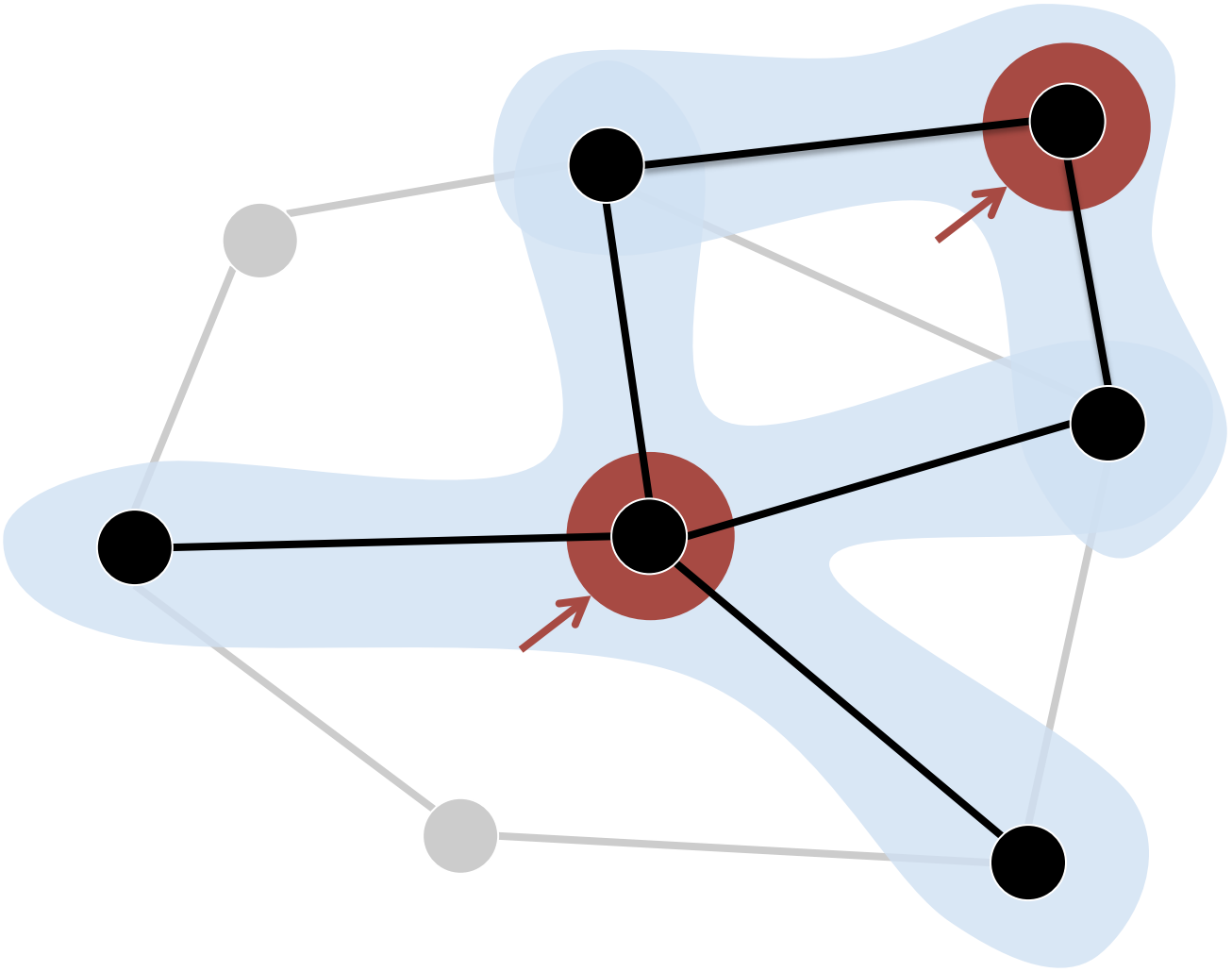
The Graph-Parallel Pattern



The Graph-Parallel Pattern



The Graph-Parallel Pattern



Many Graph-Parallel Algorithms

Collaborative Filtering
- » Alternating Least Squares
- » Stochastic Gradient Descent
- » Tensor Factorization

Structured Prediction
- » Loopy Belief Propagation
- » Max-Product Linear Programs
- » Gibbs Sampling

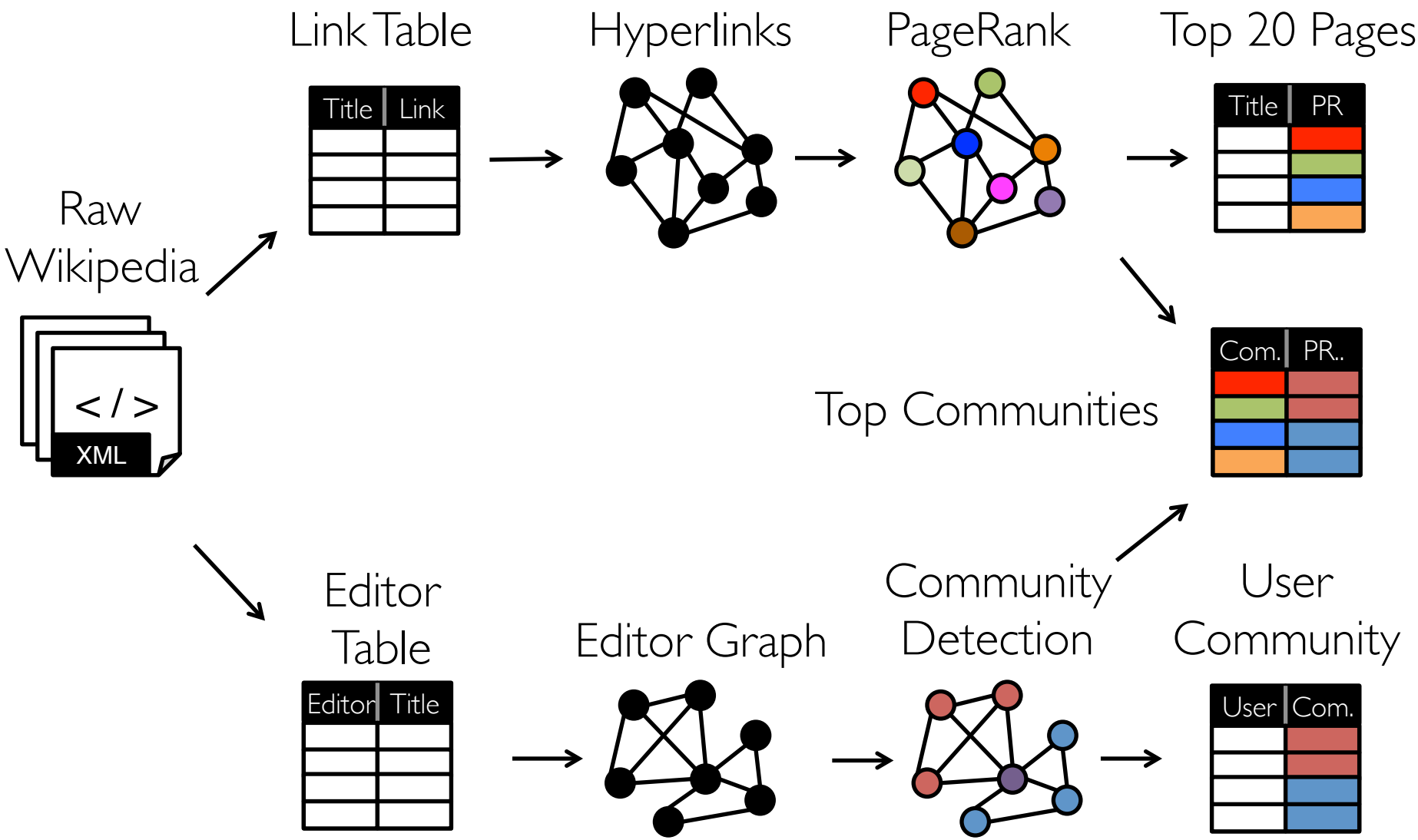
Semi-supervised ML
- » Graph SSL
- » CoEM

Community Detection
- » Triangle-Counting
- » K-core Decomposition
- » K-Truss

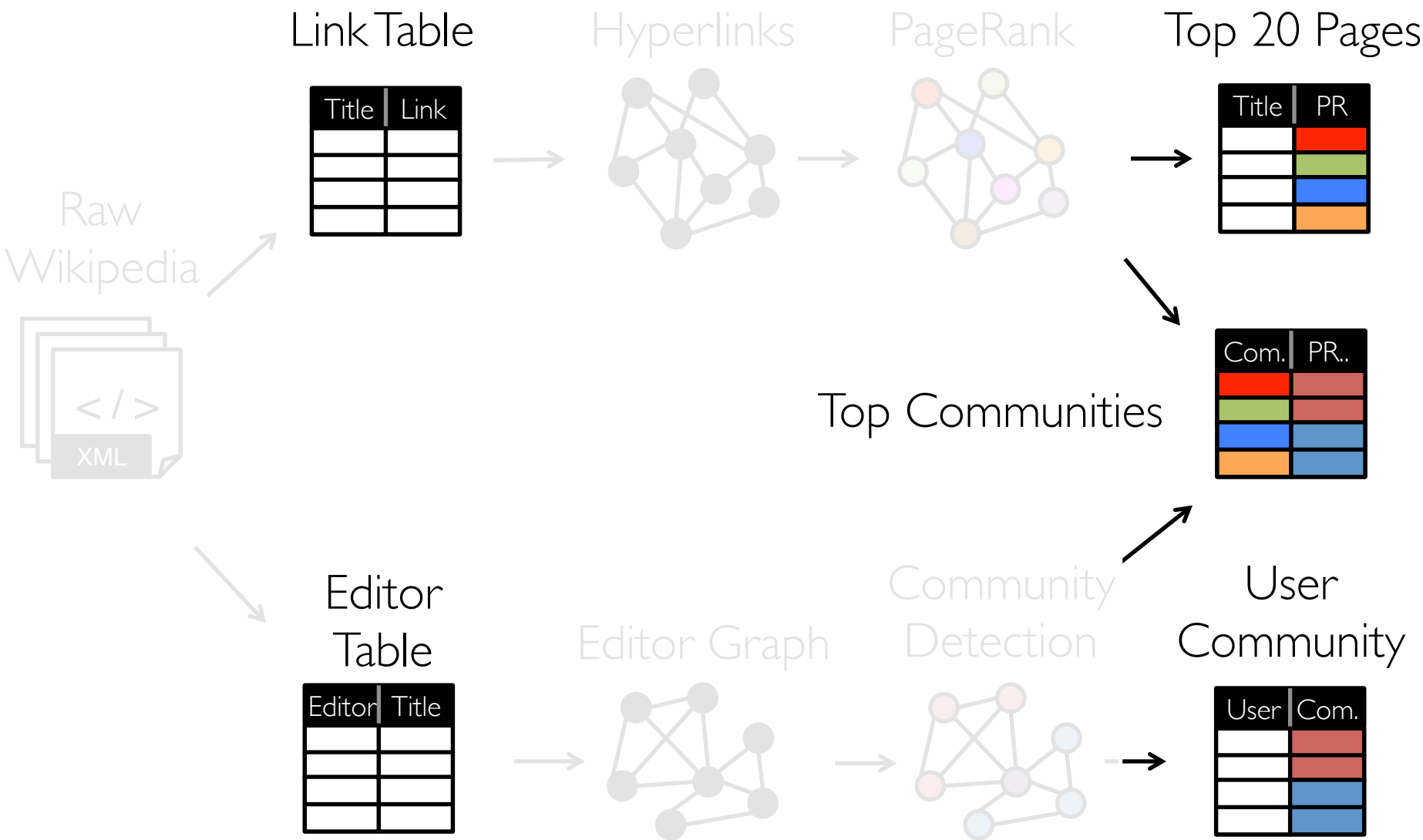
Graph Analytics
- » PageRank
- » Personalized PageRank
- » Shortest Path
- » Graph Coloring

Classification
- » Neural Networks

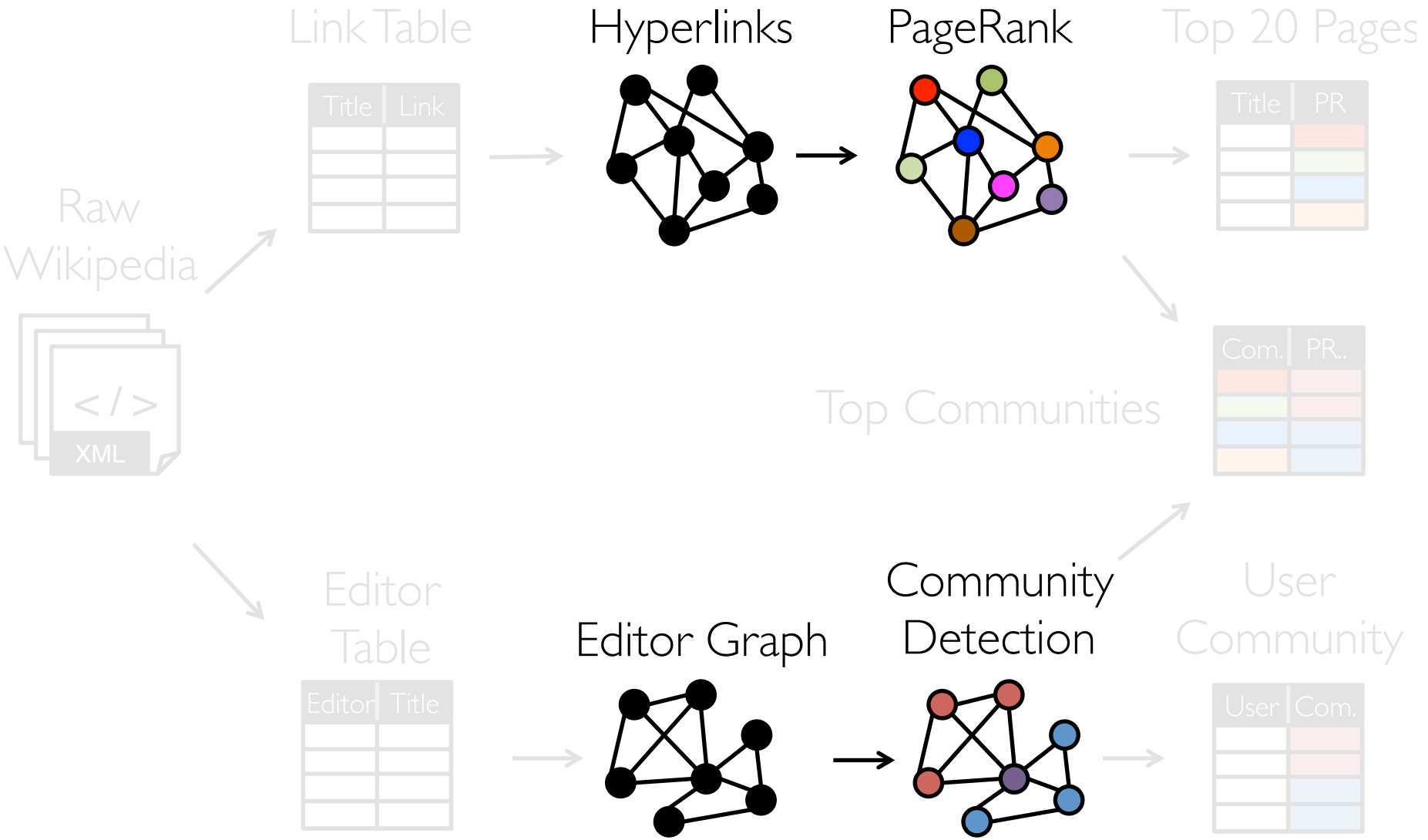
Modern Analytics



Tables


Link Table
Hyperlinks
PageRank
Top 20 Pages
Title
Link
Title
PR
Raw
Wikipedia
XML
Top Communities
Com.
PR..
Editor
Table
Editor Graph
Community
Detection
User
Community
Editor
Title
User
Com.

Graphs


Link Table
Title
Link
Hyperlinks
PageRank
Top 20 Pages
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Raw
Wikipedia
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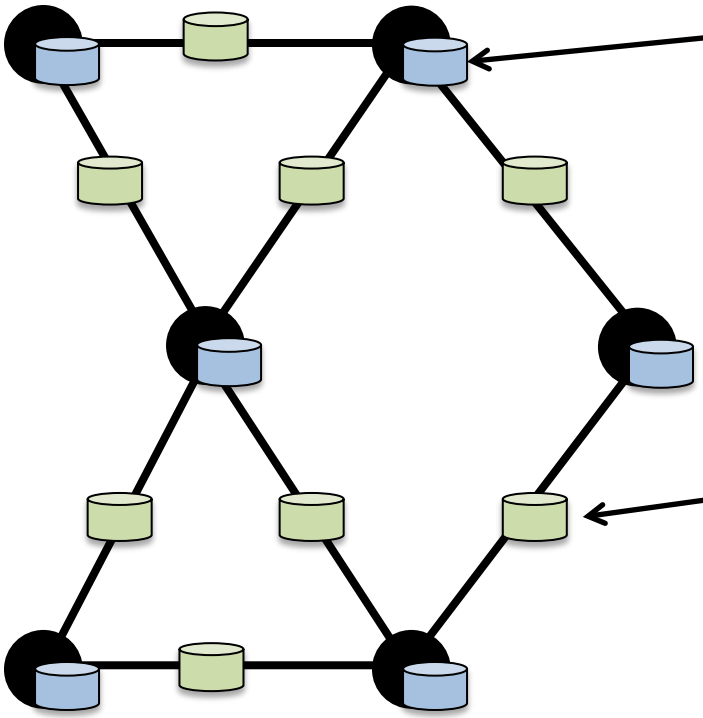
XML
Top Communities
Com.
PR..
Editor
Table
Editor
Title
Editor Graph
Community
Detection
User
Community
User
Com.

The GraphX API

Property Graphs



Vertex Property:

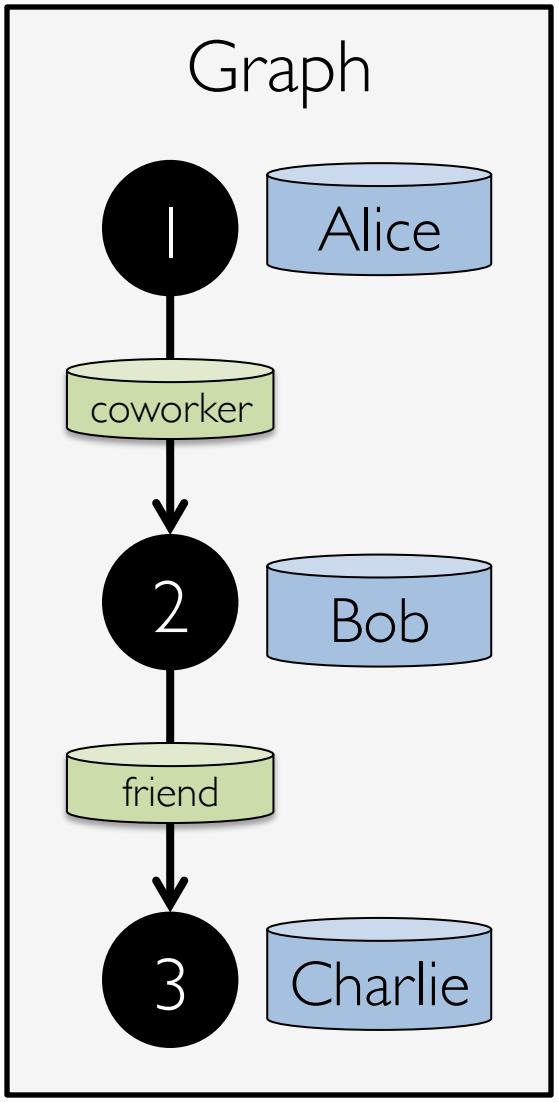
- User Profile
- Current PageRank Value

Edge Property:

- Weights
- Relationships
- Timestamps

Creating a Graph (Scala)

```
type VertexId = Long

val vertices: RDD[(VertexId, String)] =
  sc.parallelize(List(
    (1L, “Alice”),
    (2L, “Bob”),
    (3L, “Charlie”)))

class Edge[ED](
  val srcId: VertexId,
  val dstId: VertexId,
  val attr: ED)

val edges: RDD[Edge[String]] =
  sc.parallelize(List(
    Edge(1L, 2L, “coworker”),
    Edge(2L, 3L, “friend”)))

val graph = Graph(vertices, edges)
```



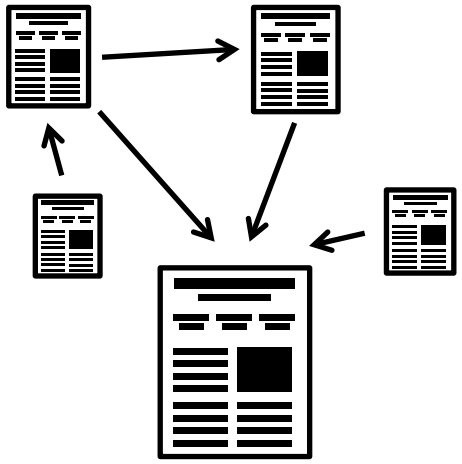
Graph Operations (Scala)

```
class Graph[VD, ED] {
    // Table Views -----------------------
    def vertices: RDD[(VertexId, VD)]
    def edges: RDD[Edge[ED]]
    def triplets: RDD[EdgeTriplet[VD, ED]]
    // Transformations ------------------------------------
    def mapVertices[VD2](f: (VertexId, VD) => VD2): Graph[VD2, ED]
    def mapEdges[ED2](f: Edge[ED] => ED2): Graph[VD2, ED]
    def reverse: Graph[VD, ED]
    def subgraph(epred: EdgeTriplet[VD, ED] => Boolean,
                 vpred: (VertexId, VD) => Boolean): Graph[VD, ED]
    // Joins ----------------------------------------
    def outerJoinVertices[U, VD2]
        (tbl: RDD[(VertexId, U)])
        (f: (VertexId, VD, Option[U]) => VD2): Graph[VD2, ED]
    // Computation ----------------------------------
    def mapReduceTriplets[A](
        sendMsg: EdgeTriplet[VD, ED] => Iterator[(VertexId, A)],
        mergeMsg: (A, A) => A): RDD[(VertexId, A)]
```

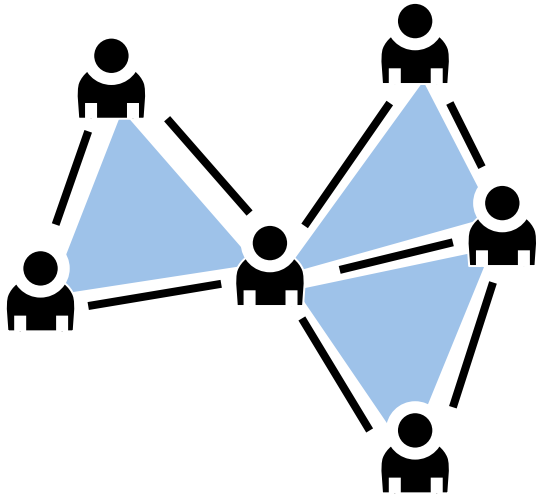
Built-in Algorithms (Scala)

```
    // Continued from previous slide
    def pageRank(tol: Double): Graph[Double, Double]
    def triangleCount(): Graph[Int, ED]
    def connectedComponents(): Graph[VertexId, ED]
    // ...and more: org.apache.spark.graphx.lib
}
```

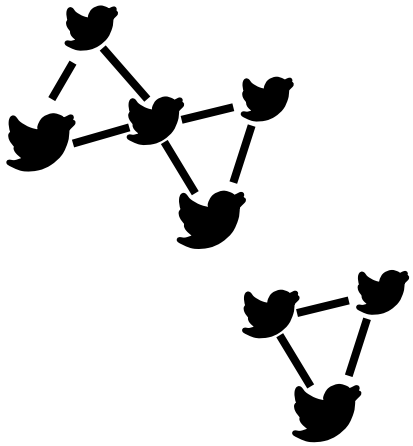
PageRank



Triangle Count

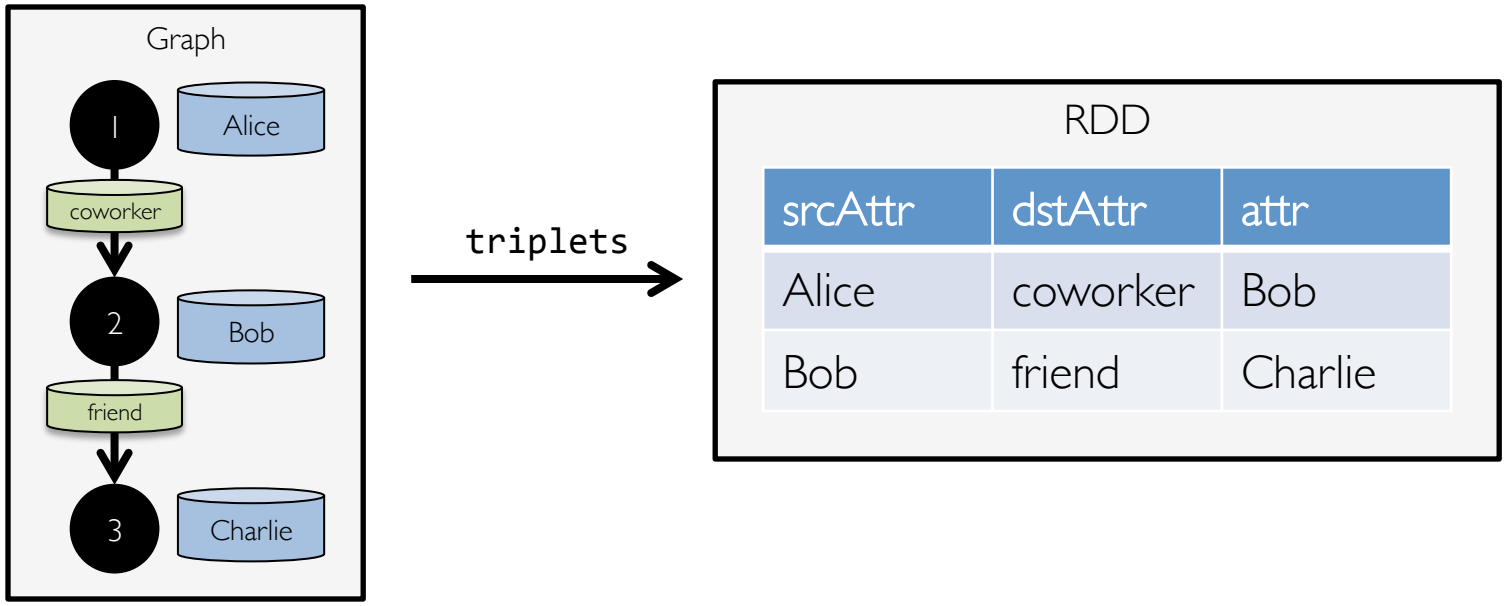


Connected Components



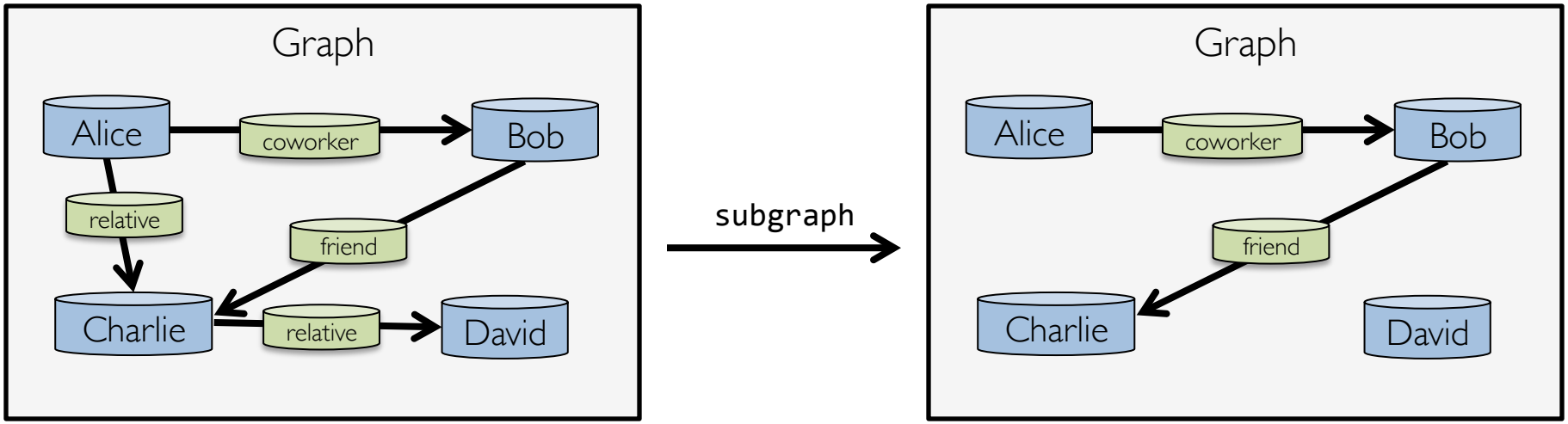
The `triplets` view

```
class Graph[VD, ED] {
    def triplets: RDD[EdgeTriplet[VD, ED]]
}

class EdgeTriplet[VD, ED](
  val srcId: VertexId, val dstId: VertexId, val attr: ED,
  val srcAttr: VD, val dstAttr: VD)
```



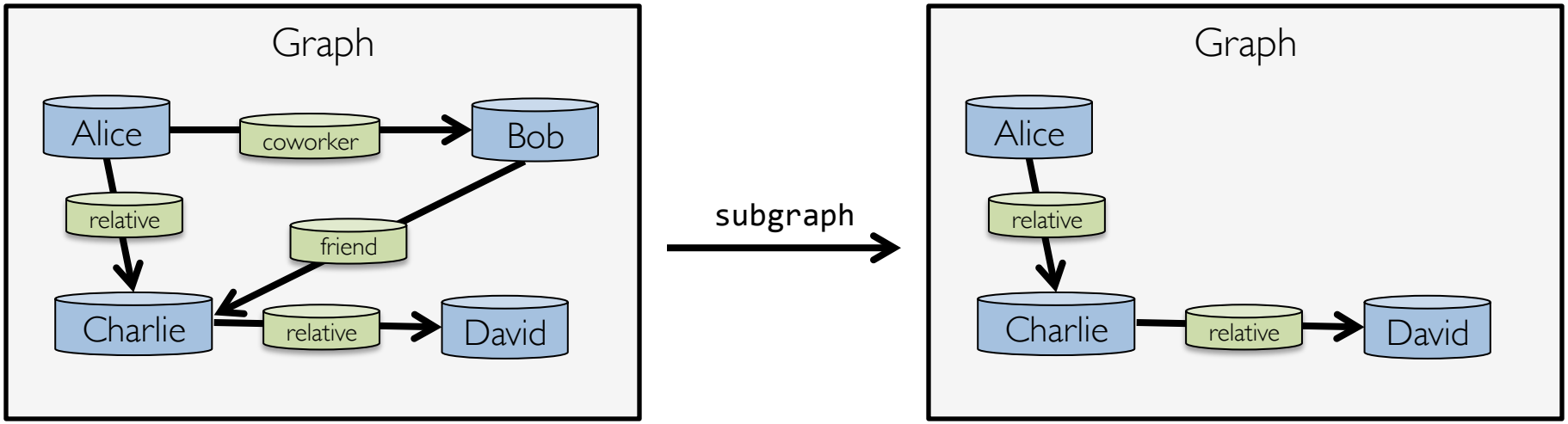
The **subgraph** transformation

```
class Graph[VD, ED] {
    def subgraph(epred: EdgeTriplet[VD, ED] => Boolean,
                 vpred: (VertexId, VD) => Boolean): Graph[VD, ED]
}

graph.subgraph(epred = (edge) => edge.attr != “relative”)
```



The subgraph transformation

```
class Graph[VD, ED] {
    def subgraph(epred: EdgeTriplet[VD, ED] => Boolean,
                 vpred: (VertexId, VD) => Boolean): Graph[VD, ED]
}

graph.subgraph(vpred = (id, name) => name != "Bob")
```

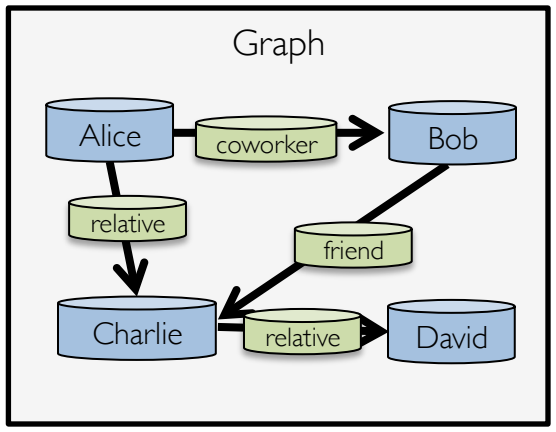


Computation with `mapReduceTriplets`

```
class Graph[VD, ED] {
  def mapReduceTriplets[A](
    sendMsg: EdgeTriplet[VD, ED] => Iterator[(VertexId, A)],
    mergeMsg: (A, A) => A): RDD[(VertexId, A)]
}
```

upgrade to **`aggregateMessages`** in Spark 1.2.0

```
graph.mapReduceTriplets(
  edge => Iterator(
    (edge.srcId, 1),
    (edge.dstId, 1)),
  _ + _)
```



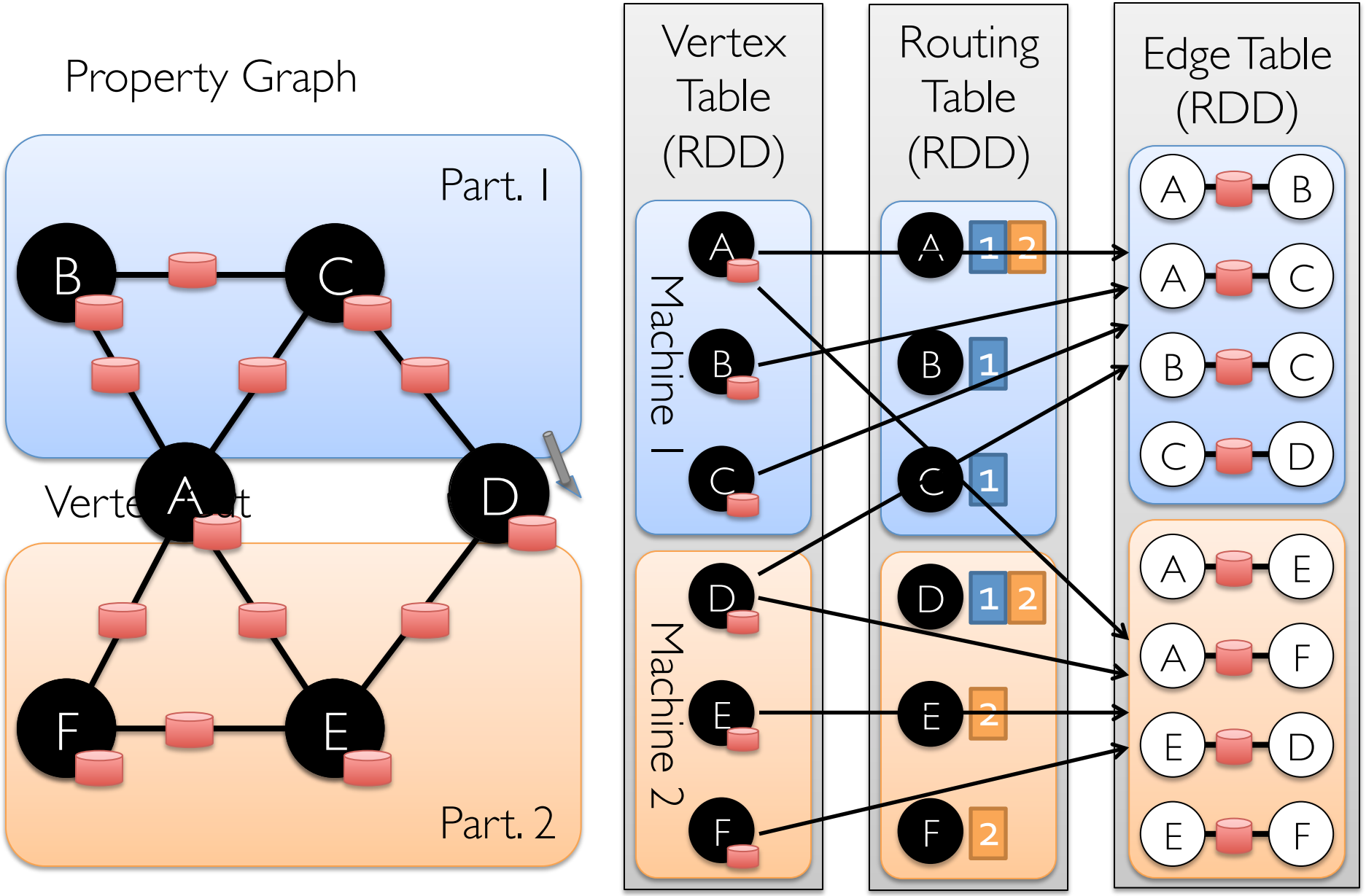
`mapReduceTriplets`

RDD

vertex id	degree
Alice	2
Bob	2
Charlie	3
David	1

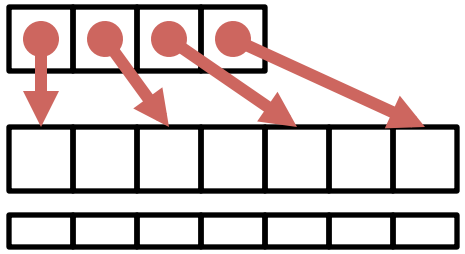
How GraphX Works

Encoding Property Graphs as RDDs

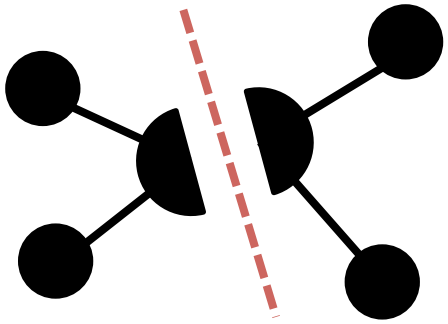


Graph System Optimizations

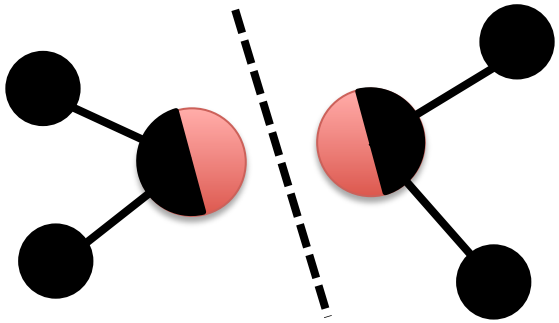
Specialized Data-Structures



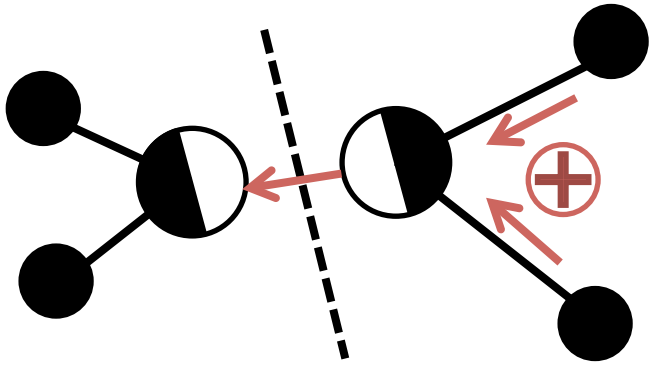
Vertex-Cuts Partitioning



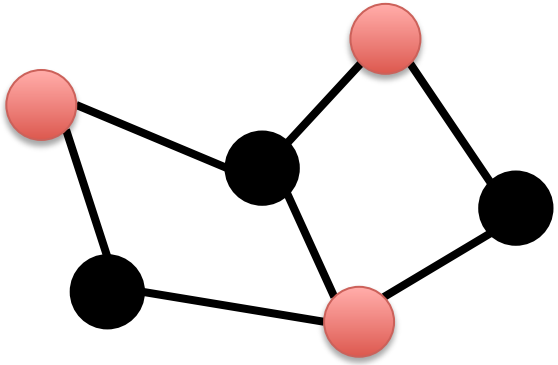
Remote Caching / Mirroring



Message Combiners

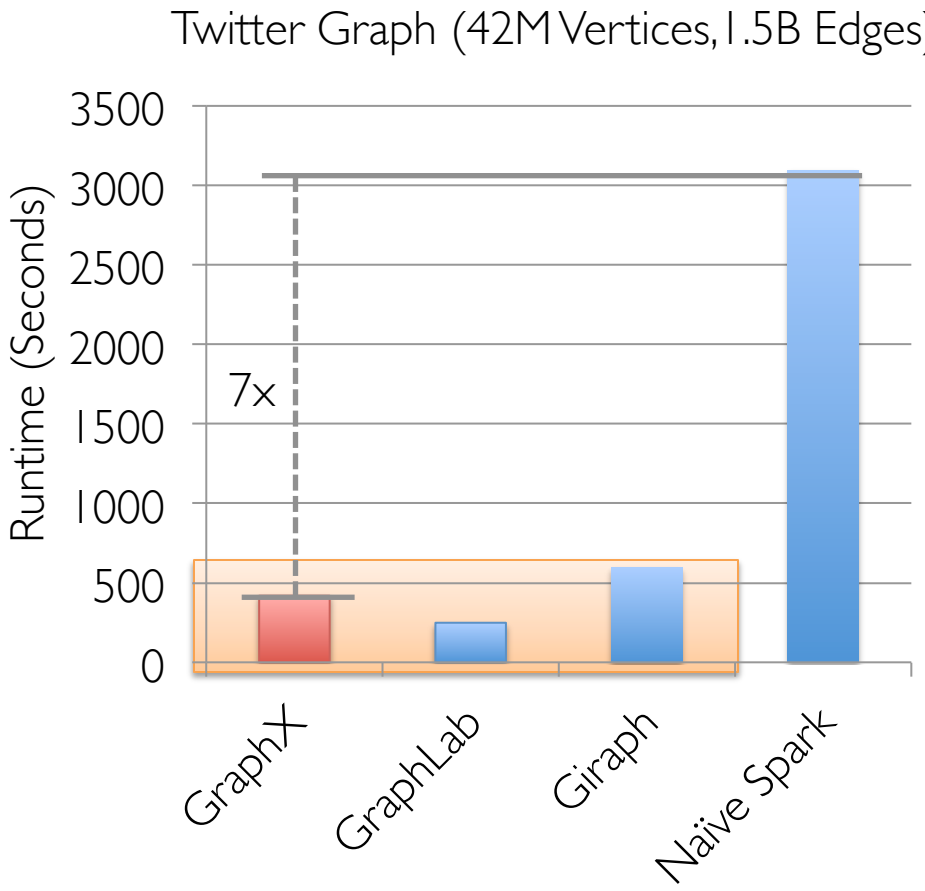


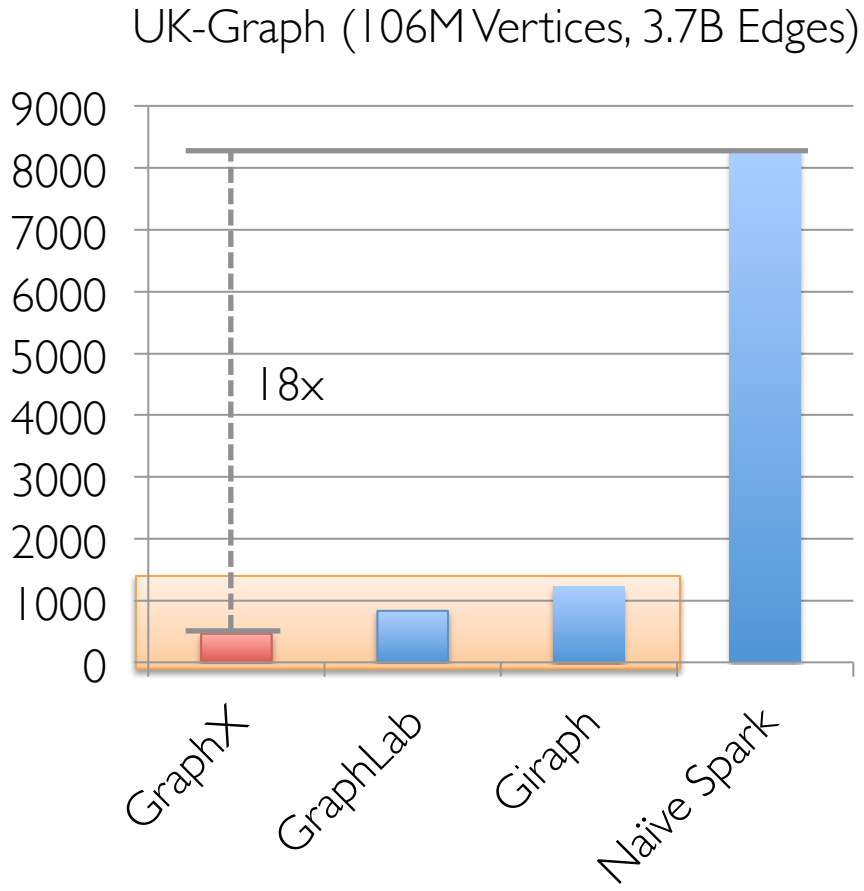
Active Set Tracking



PageRank Benchmark

EC2 Cluster of 16 x m2.4xLarge (8 cores) + 1GigE





GraphX performs comparably to
state-of-the-art graph processing systems.

Future of GraphX

1. Language support
   a) Java API: PR #3234
   b) Python API: collaborating with Intel, SPARK-3789

2. More algorithms
   a) LDA (topic modeling): PR #2388
   b) Correlation clustering
   c) Your algorithm here?

3. Speculative
   a) Streaming/time-varying graphs
   b) Graph database–like queries

Thanks!

http://spark.apache.org/graphx

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crankshaw@eecs.berkeley.edu