

Spark SMTI

Yilong Geng,  
Mingyu Gao

Background

GaleShapley  
Algorithm

Distributed  
SMTI

Implementation  
on Spark

Performance

End

# Distributed Stable Marriage with Incomplete List and Ties using Spark

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# Overview

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# Marriage Problem

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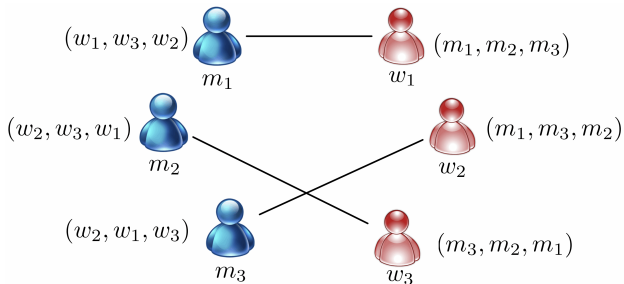
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# Stable Marriage with Incomplete list and Ties (SMTI)

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## Blocking Pair

A pair  $mw$  is blocking  $M$ , if  $mw \in E \setminus M$  (they are an acceptable pair and they are not matched) and

- $w$  is not engaged or strictly prefers  $m$  to her fiancé, **and**
- $m$  is not engaged or strictly prefers  $w$  to his fiancé.

## Stable

A marriage is called stable if there is no blocking pair.

## Incomplete list and Ties

Preference list can be incomplete, and/or have ties

**Our goal:** find a stable marriage with maximum marriage size

# Basic GS Algorithm

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```
1: Initialize all  $m$  and  $w$  to free
2: while there is a free  $m$  who has non-empty list do
3:    $w =$  top in the list
4:   if  $w$  is free then
5:      $mw$  engage
6:   else
7:     Some  $m'w$  already engaged
8:     if  $w$  prefers  $m$  to  $m'$  then
9:        $mw$  engage,  $m'$  becomes free
10:    Remove  $w$  from  $m'$ 's list
11:   else
12:      $m'w$  remain engaged, remove  $w$  from  $m$ 's list
13:   end if
14: end if
15: end while
```

---

- Single machine complexity:  $O(e)$
- Result size  $\geq \frac{1}{2}$  of the maximum size

# Modified GS for SMTI

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- Zoltan Kiraly, *Linear Time Local Approximation Algorithm for Maximum Stable Marriage.*
- Key modification
  - Men will go through his list twice
  - Break tie with the current status of the men and women
- Single machine complexity:  $O(e)$
- Result size  $\geq \frac{2}{3}$  of the maximum size

# Distributed Modified GS

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- Preference list length is usually much shorter than  $n$ .
- Proposers (men) and acceptors (women) are represented in RDD, including their preference lists.
- In each iteration, all proposers will propose simultaneously, and acceptors pick the best proposals.
- Communication complexity:  $O(n)$  per iteration.
- Single worker reduce complexity: maximum list length.

# Use Pregel?

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- Two groups of people form a (sparse) bipartite graph, with (incomplete) preference lists as edges
- Advantages
  - Pregel handles partitioning better
  - Simple API
- Problems
  - Different attributes for proposers/acceptors
  - Each vertex only sends message along one edge in each iteration
  - Two rounds of message transfer in one iteration



# Pseudocode

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```
do {  
    // Proposers --> acceptors  
    val proposals = proposers.filter( isActive )  
        .map( makeProposal ).groupByKey()  
        .cache()  
  
    acceptors = acceptors.leftOuterJoin(proposals)  
        .mapValues( handleProposal )  
        .cache()  
  
    // Acceptors --> proposers  
    val responses = acceptors  
        .map( makeResponse )  
        .cache()  
  
    proposers = proposers.leftOuterJoin(responses)  
        .mapValues( handleResponse )  
        .cache()  
  
} while ( hasActiveProposers )
```

# Performance Scaling

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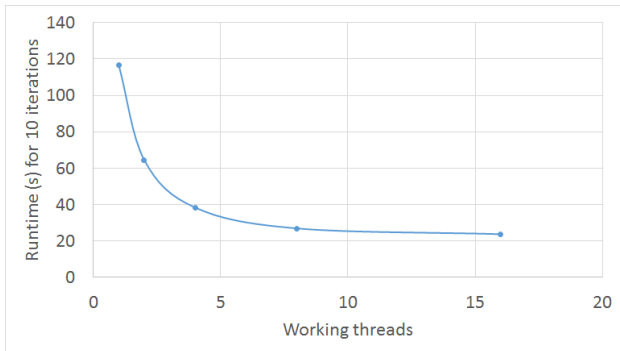
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With fixed number of RDD partitions (64 partitions)

# Performance Scaling

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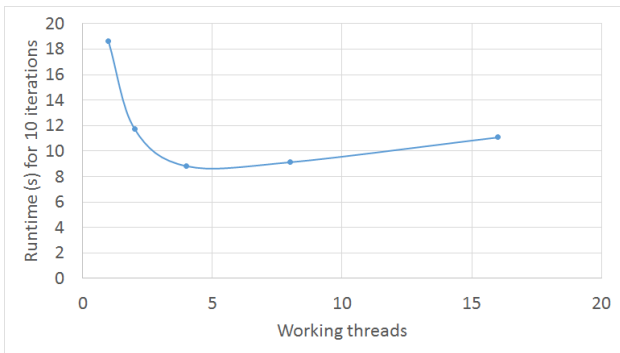
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With number of RDD partitions equal to number of threads

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# Thanks!

<https://github.com/gaomy3832/spark-smti>