

# Coalition Formation towards Energy-Efficient Collaborative Mobile Computing

---

**Liyao Xiang**, Baochun Li, Bo Li  
Aug. 3, 2015

# Collaborative Mobile Computing

---

- ▶ **Mobile offloading**: migrating the computation-intensive portion of an app to the cloud to execute.
- ▶ **Gain**: trades the relatively low communication energy expense for high computation power consumption.
- ▶ **Loss**: suffers high network latency.
- ▶ New features such as *Continuity* made offloading tasks to nearby devices possible.

# Coalition Formation of Mobile Users

---

- ▶ Previous works assume fully cooperative mobile users.
- ▶ We assume users are:
  - ▶ cooperative: collaborates under agreements.
  - ▶ individually rational: prefers coalition if it benefits.
- ▶ We study the problem of coalition formation among a group of mobile users targeting at the same job.

# Coalition Formation of Mobile Users

---

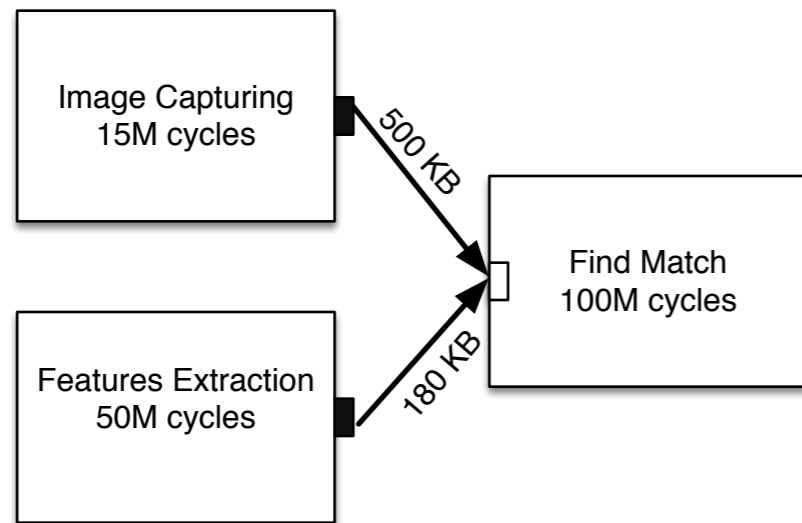
- ▶ User case: crowdsourcing, content sharing, indoor localization, etc.
- ▶ Key questions:
  - ▶ Given a job partitioned into several tasks, how does a group of users form coalitions?
  - ▶ Within each coalition, how to distribute the tasks to each user?

# System Model

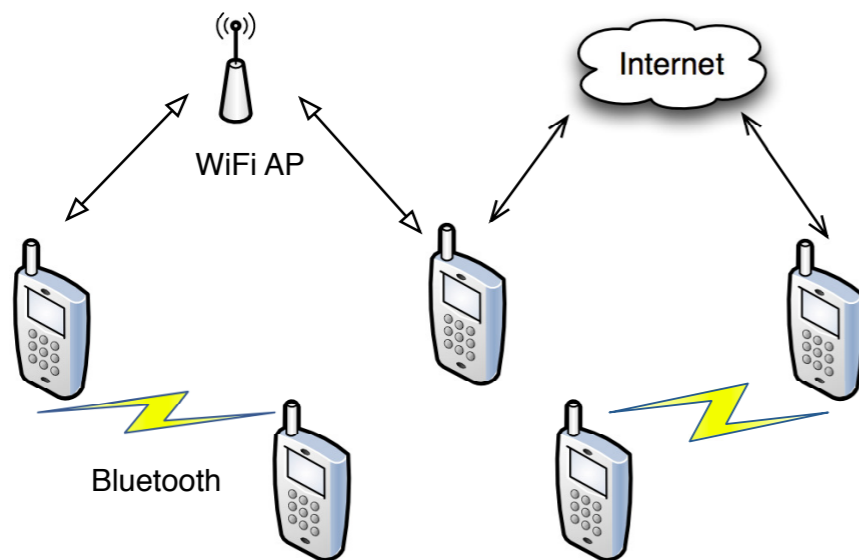
---

- ▶ **A centralized approach:** an arbitrator profiles user's info, organizes users into groups, and assigns tasks to each group.
- ▶ **A distributed scheme:** mobile users exchange profiles with users targeting at the same job. Based on the estimated energy cost, users decide to merge into one group or split up.
- ▶ A profile is generated by program static analysis tools.

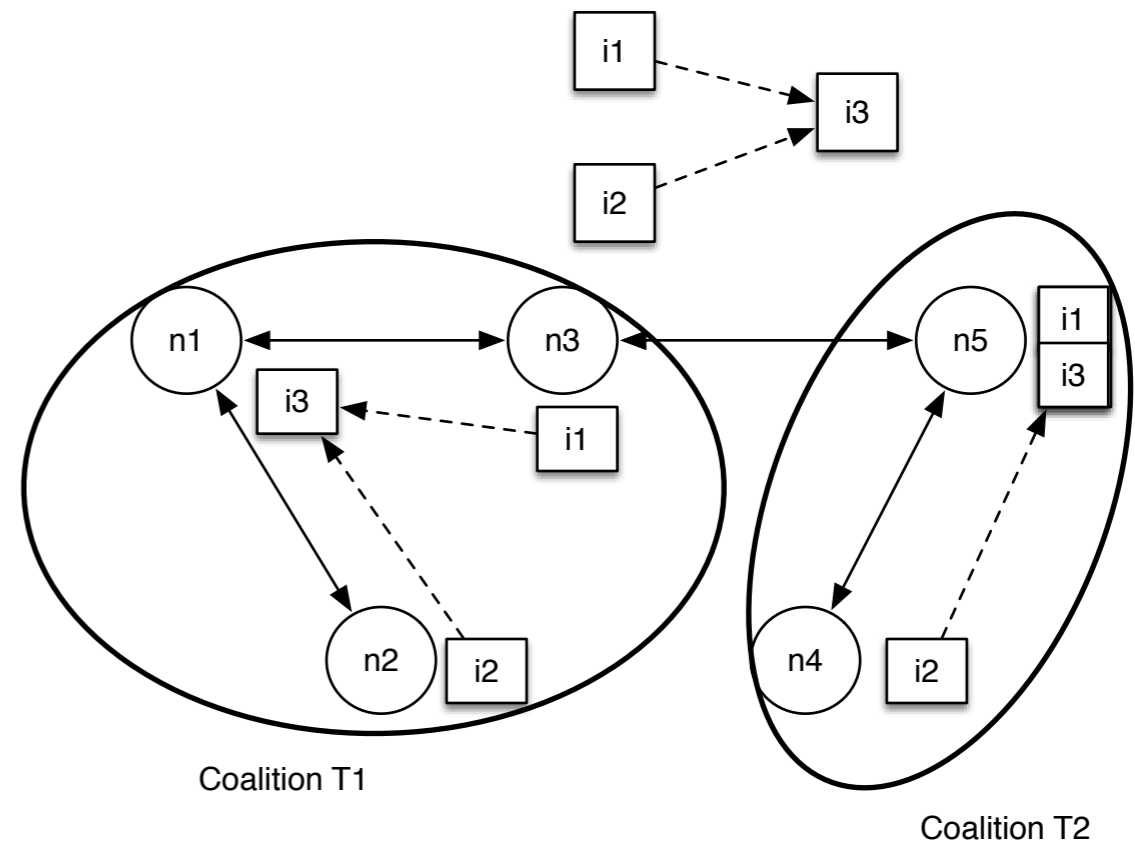
# System Model



Task graph



Resource graph



An example of mapping tasks to a set of devices

# Task Distribution

---

- ▶ Objective: minimizing the overall energy expense over all partitions of the resource graph with placement constraints.
- ▶  $\mathcal{B}$  is the set of all partitions.  $T$  represents one coalition.  $C(T)$  is the sum of the energy expense on all mobile devices in coalition  $T$ .

$$\min_{\mathcal{P} \in \mathcal{B}} \sum_{T \in \mathcal{P}} \min C(T).$$

# Task Distribution

---

- ▶ To assign the binary variable  $s_{i,n}$  representing task  $i$  is to be executed on device  $n$ .
- ▶ Placement constraints:

$$\sum_{n \in T} s_{i,n} = 1, \forall i \in \mathbb{V}, \quad (5)$$

$$\sum_{i \in \mathbb{V}} s_{i,n} \geq 1, \forall n \in T, \quad (6)$$

$$s_{i,n} s_{j,m} e_{i,j} \leq l_{n,m}, \forall i \neq j, \forall n \neq m, n, m \in T \quad (7)$$

$$s_{i,n} \leq r_{i,n}, \forall i, \forall n \in T, \quad (8)$$

$$s_{i,n} \in \{0, 1\}, \forall n \in T, \forall i \in \mathbb{V}. \quad (9)$$



# Coalition Formation

---

- ▶ The centralized approach is non-convex and NP-hard. How about going distributed?
- ▶ Collaboration among mobile users is modelled as a **non-transferrable utility coalition game**  $(N, v)$  where  $N$  is the entire set of users, and  $v$  is the utility for the coalition which is defined as the negative energy cost.

- ▶ Partition:

**Definition 1:** A *collection* is any family  $T = \{T_1, \dots, T_l\}$  of mutually disjoint coalitions. If additionally  $\bigcup_{j=1}^l T_j = N$ , the collection  $T$  is called a *partition* of  $N$ .

# Coalition Formation

---

- ▶ Comparison relation:

**Definition 2:** Assume  $A$  and  $B$  are partitions of the same set  $C$ , a *comparison relation*  $\triangleright$  is defined as,  $A \triangleright B$  means that the way  $A$  partitions  $C$  is preferable to the way  $B$  partitions  $C$ .

- ▶ Pareto order: the transformation of coalitions through Pareto order can only happen when it at least **strictly improves** the utility of one user, i.e., given two partitions  $T$  and  $T'$ , with  $\phi(T)$  representing the energy cost of  $T$ , the comparison relation is expressed as:

$$T \triangleright T' \iff \forall n, \phi_n(T) \leq \phi_n(T') \text{ and } \exists m, \phi_m(T) < \phi_m(T')$$

# Coalition Formation

---

- ▶ Two rules to transform coalitions:

**Merge:**  $\{T_1, \dots, T_k\} \cup P \rightarrow \{\bigcup_{j=1}^k T_j\} \cup P$ , where  $\{\bigcup_{j=1}^k T_j\} \triangleright \{T_1, \dots, T_k\}$ .

**Split:**  $\{\bigcup_{j=1}^k T_j\} \cup P \rightarrow \{T_1, \dots, T_k\} \cup P$ , where  $\{T_1, \dots, T_k\} \triangleright \{\bigcup_{j=1}^k T_j\}$ .

- ▶ Based on the above rules, we derive the algorithm:

---

**Algorithm 1** Collaborative Computing Game through Merge and Split

---

Input: Initial partition  $T = \{T_1, \dots, T_l\} = \mathbb{N}$

Output: Final partition  $T^{final}$

**repeat**

$T = \text{Merge}(T)$ ;

$T = \text{Split}(T)$ ;

**until** merge and split terminates.

$T^{final} = T$ .

---

# Stability Analysis

---

- ▶ Definition: we consider a partition  $T$  is stable if for any collection  $C$  of the entire user set  $N$  that

$$C[T] \triangleright C, \quad (14)$$

where

$$C[T] = \{T_1 \cap \bigcup\{C\}, \dots, T_k \cap \bigcup\{C\}\} \setminus \{\emptyset\}.$$

- ▶ We prove that the stability defined above implies **contractually individual stability**, i.e., a state that no player can benefit from moving its coalition to another without making others worse off.

# Dc-Stable

---

- ▶ We proved our merge-and-split mechanism is stable if allowing users to transfer between coalitions by merge and split. The stable partition is called Dc-stable partition.
- ▶ If a Dc-stable partition  $T$  exists, then  $T$  is the **unique outcome** of every iteration of merge and split.

# Performance Evaluation

---

- ▶ Setup

- ▶ Computation cycles of each task is 20-100 M cycles.
- ▶ Data transferred is 10-1000 KB on each link.
- ▶ Energy consumption in data transmission is 20-200mJ/KB.
- ▶ Computation energy cost is 40-60 mJ/M cycles.

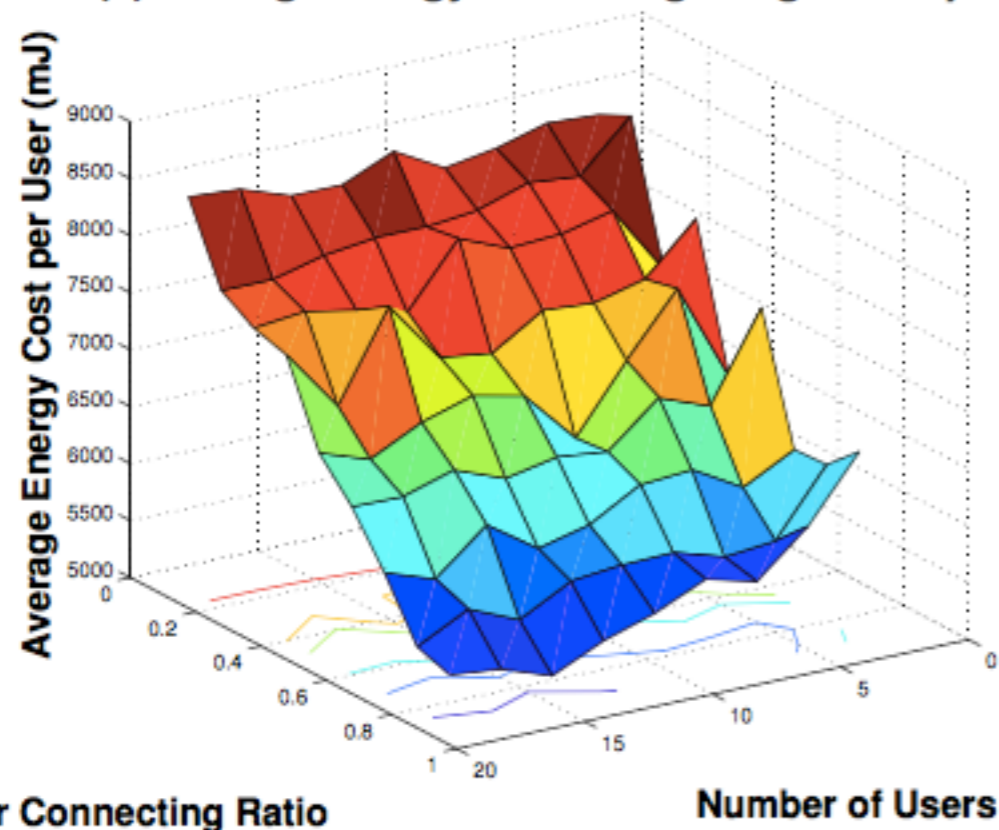
# Performance Evaluation

## ► Average Energy Cost

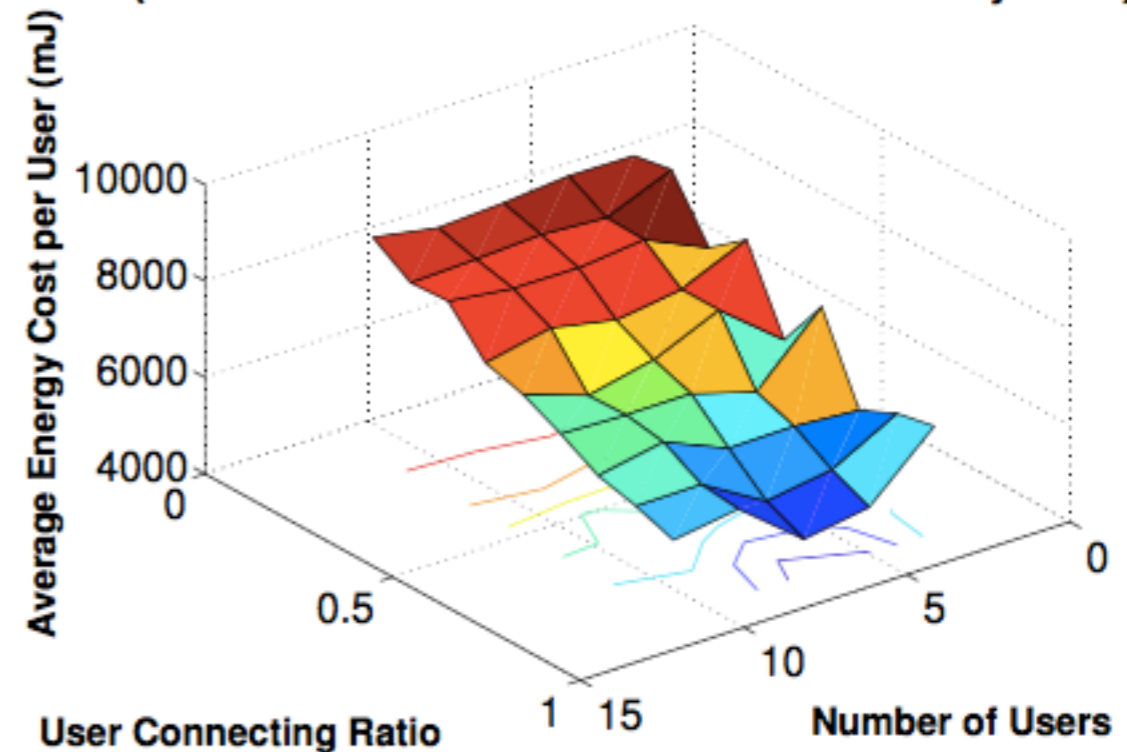
AVERAGE ENERGY COST PER USER OVER ALL CASES

	Non-coop	Centralized	Merge & Split
Energy Cost(J)	8.49	6.86	6.97

(a) Average energy cost using merge and split.

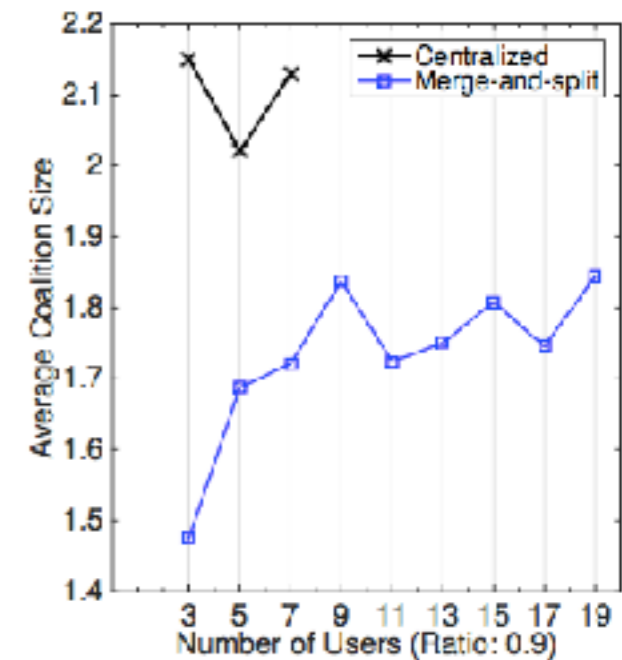
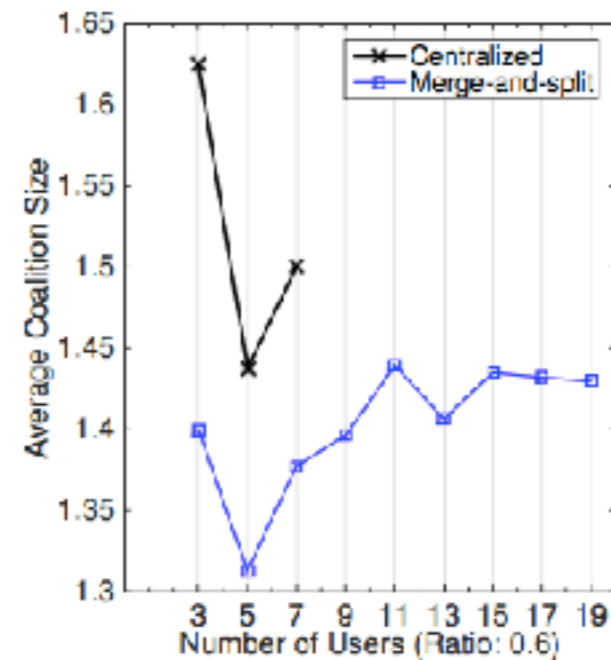
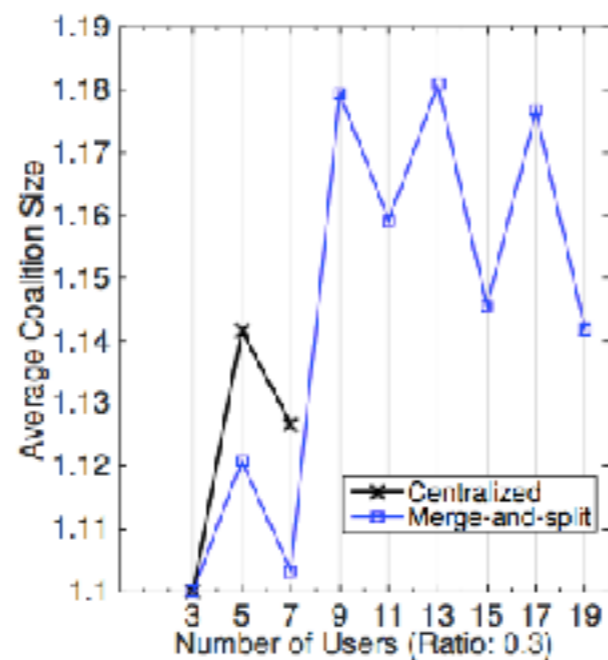
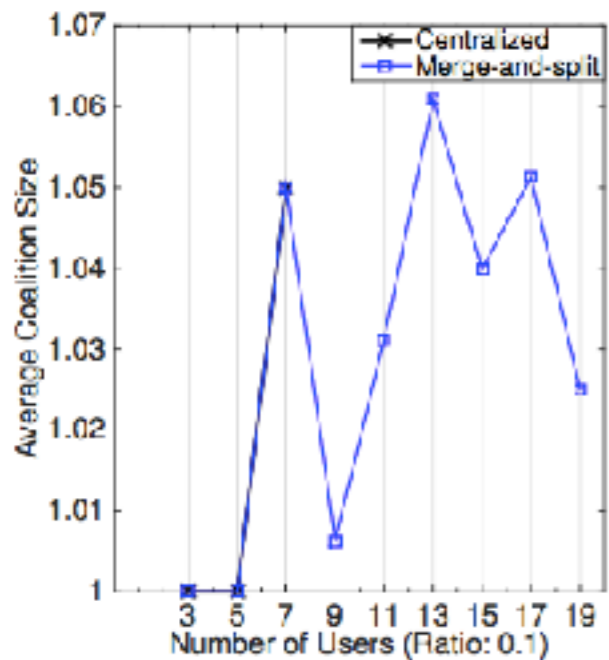


(b) Average energy cost using centralized algorithm (intractable when the number of users is beyond 7).



# Performance Evaluation

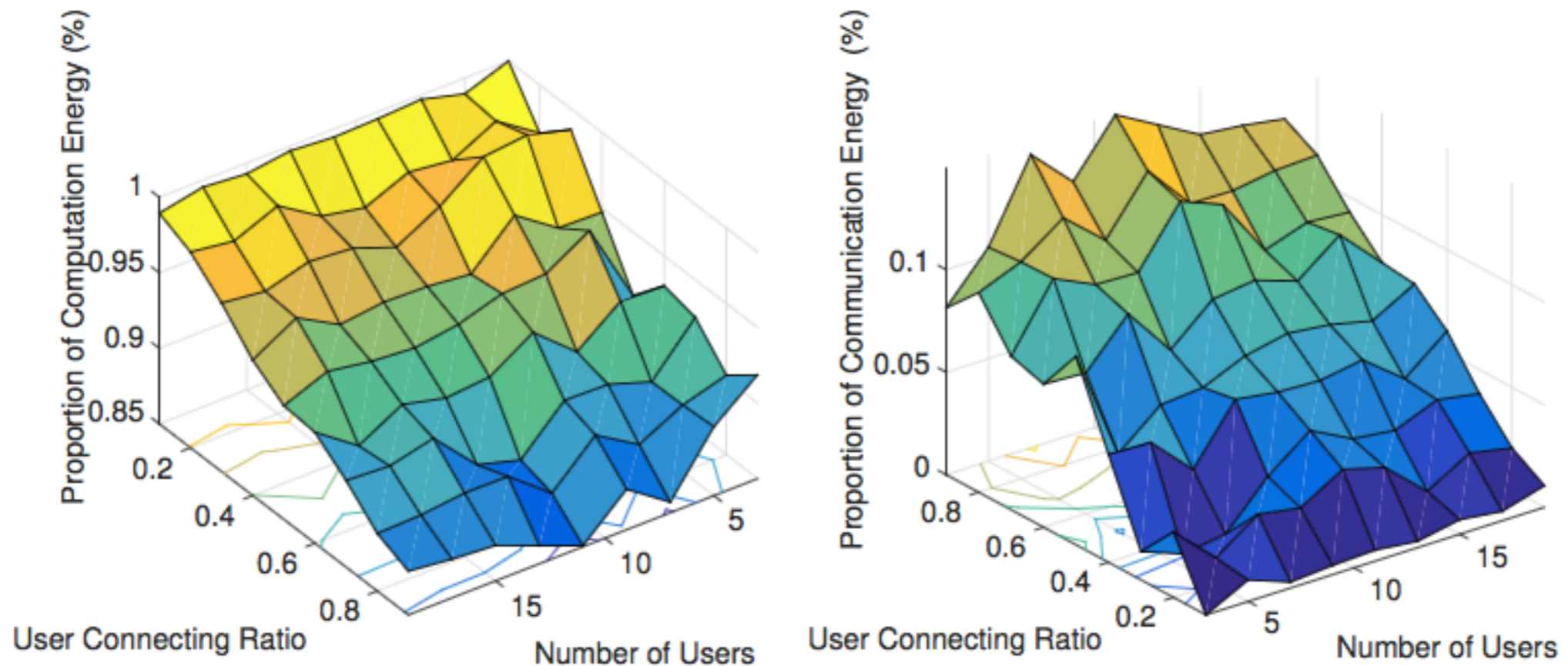
- ▶ Average coalition size.





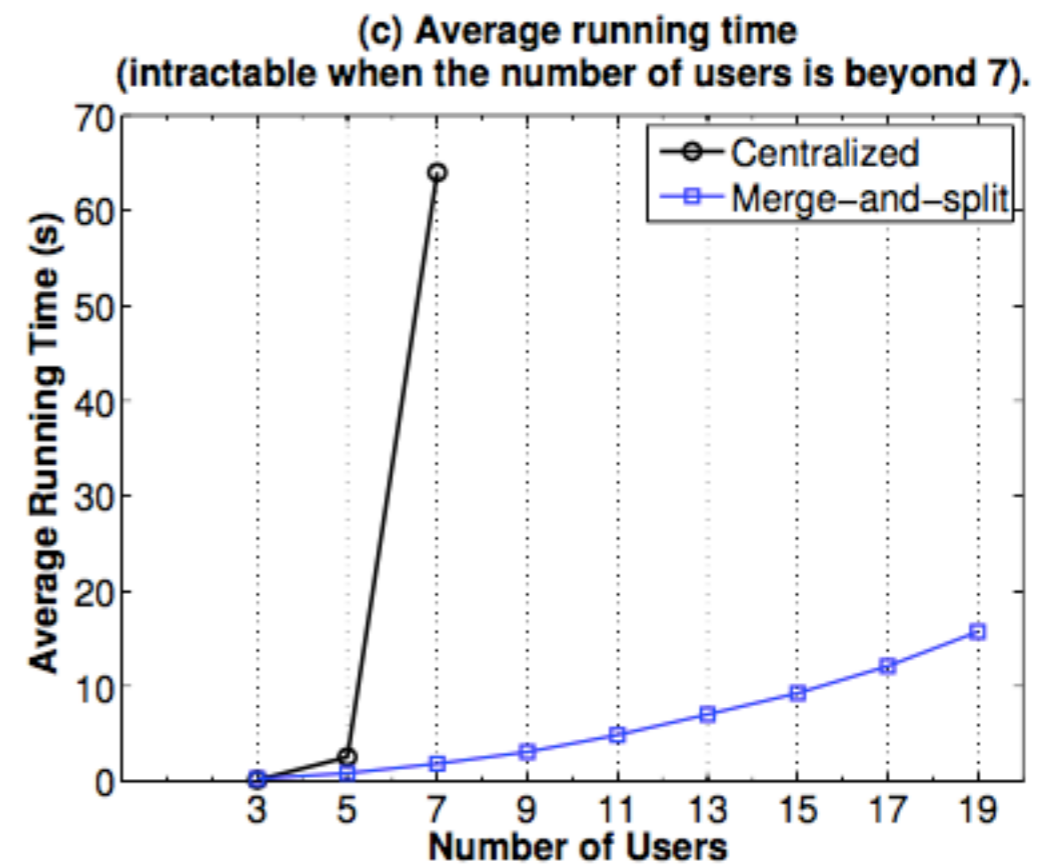
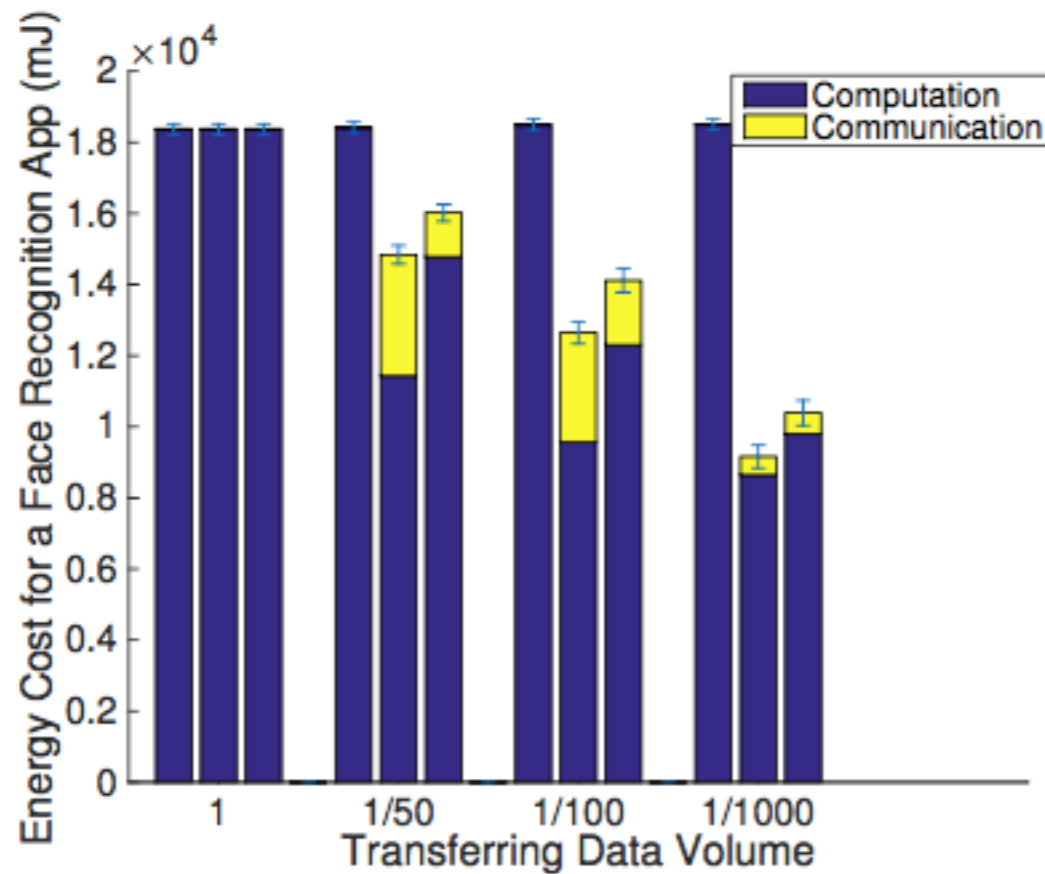
# Performance Evaluation

- ▶ Average proportion of computation and communication cost.



# Performance Evaluation

- ▶ Emulation for a real-world app & running time comparison.



# Conclusion

---

- ▶ We formulate the task assignment problem as a 0-1 integer programming problem and use heuristic method to solve it.
- ▶ We devise a distributed merge-and-split algorithm to allow collaborative and individually rational users to form coalitions.
- ▶ We reveal the conditions under which the scheme yields a stable partition.

Q & A.

Thank you.