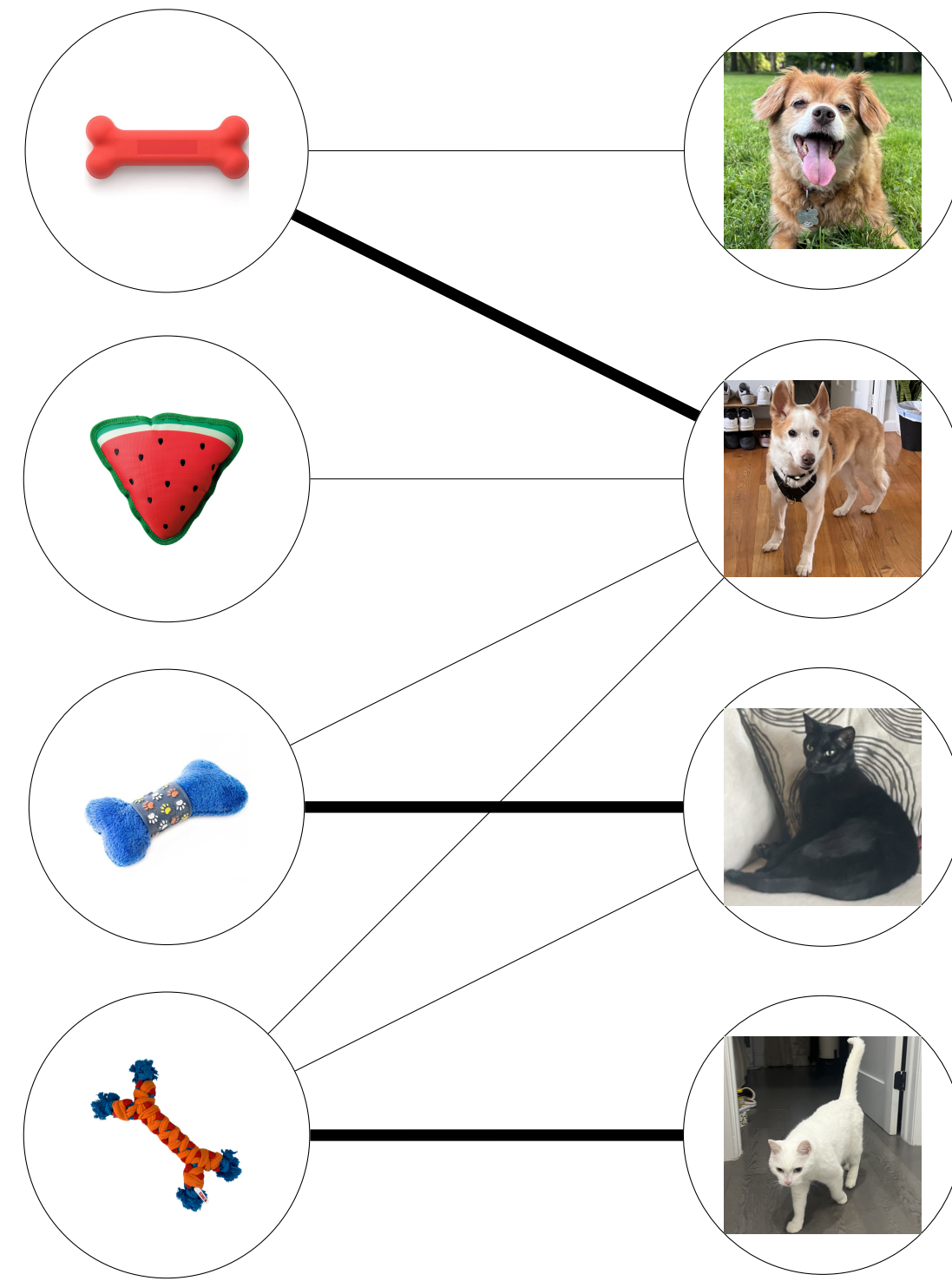


Online Matching

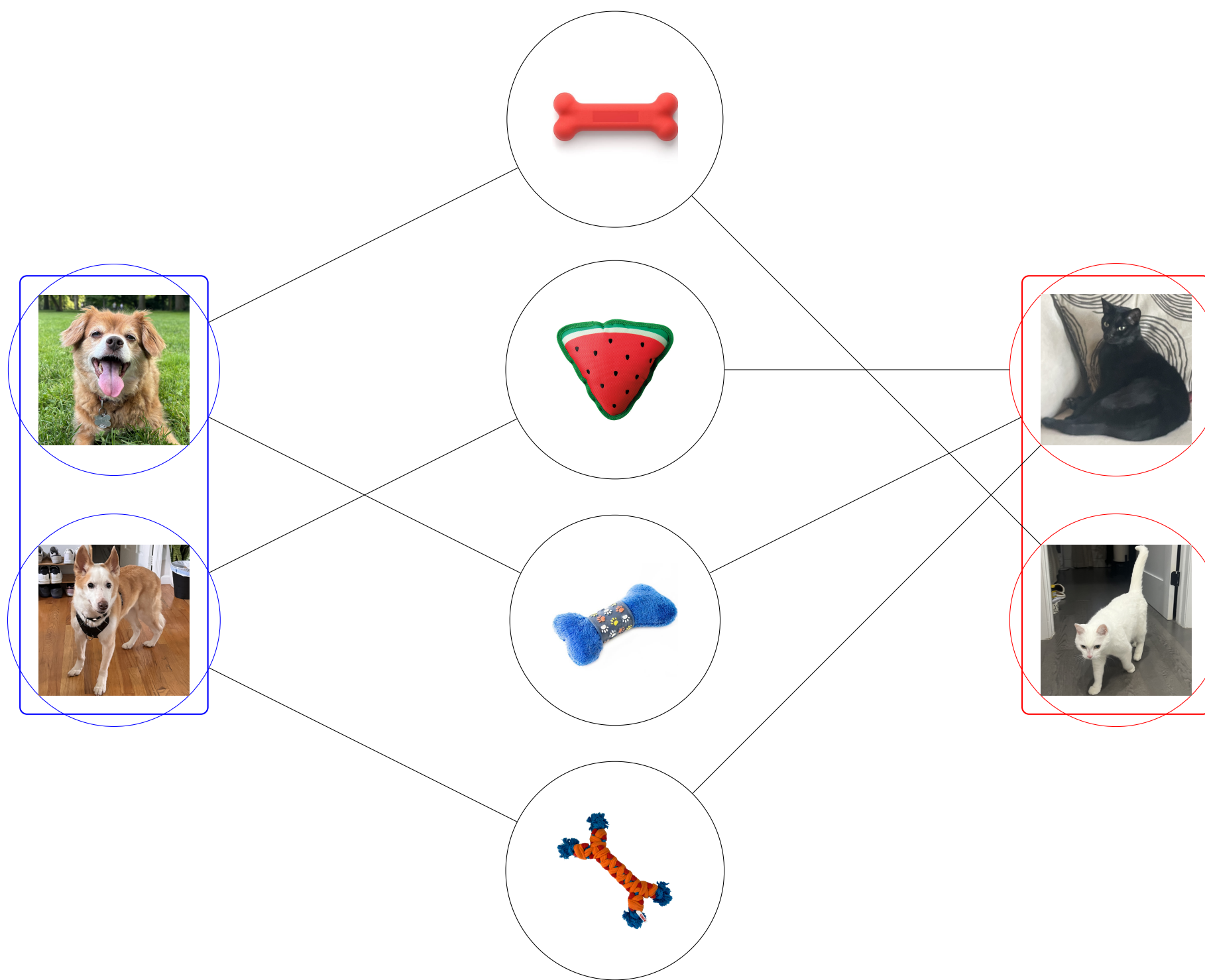
The online matching problem is a fundamental challenge in algorithm design where vertices from one set (e.g., buyers, jobs) arrive sequentially and must be matched to a fixed set of offline vertices (e.g., items, workers) without future knowledge. Decisions are irrevocable, requiring algorithms to optimize objectives such as maximizing total weight or ensuring fairness under constraints.



No randomized algorithm can achieve a better competitive ratio of $1 - 1/e$ in expectation. The RANKING algorithm gives this guarantee.

Online Fair Matching

In real-world scenarios, fairness and diversity often require considering "group" or "class" constraints.



Main Contributions

- **Impossibility Result 1:** No non-wasteful algorithm can achieve α -CEF in expectation for $\alpha > \frac{e^2-1}{e^2+1}$, or $1 - 1/e$ -CMMS.
- **Impossibility Result 2:** For any $\varepsilon > 0$, there exists a problem instance such that no (possibly randomized) non-wasteful online algorithm with an α -CEF guarantee can achieve an approximation to the USW objective greater than $\frac{1}{1+\alpha} + \varepsilon$.
- **Impossibility Result 3:** No non-wasteful *divisible* algorithm can achieve α -CEF for $\alpha > 0.677$.
- **Algorithmic Results:** Non-wasteful random assignment satisfies non-wastefulness, $1/2$ -CEF, $1/2$ -CMMS, and $1/2$ -USW.

Problem Formulation

We study online matching in a bipartite graph $G = (N, M, E)$, where N (agents) is partitioned into k known classes N_1, \dots, N_k , and M represents items [2]. Agents $a \in N$ and items $o \in M$ are adjacent if $(a, o) \in E$. A matching is represented by $X = (x_{a,o})$, where $x_{a,o} \in \{0, 1\}$ indicates whether o is matched to a . For divisible matchings, $x_{a,o} \in [0, 1]$.

Fairness Definitions:

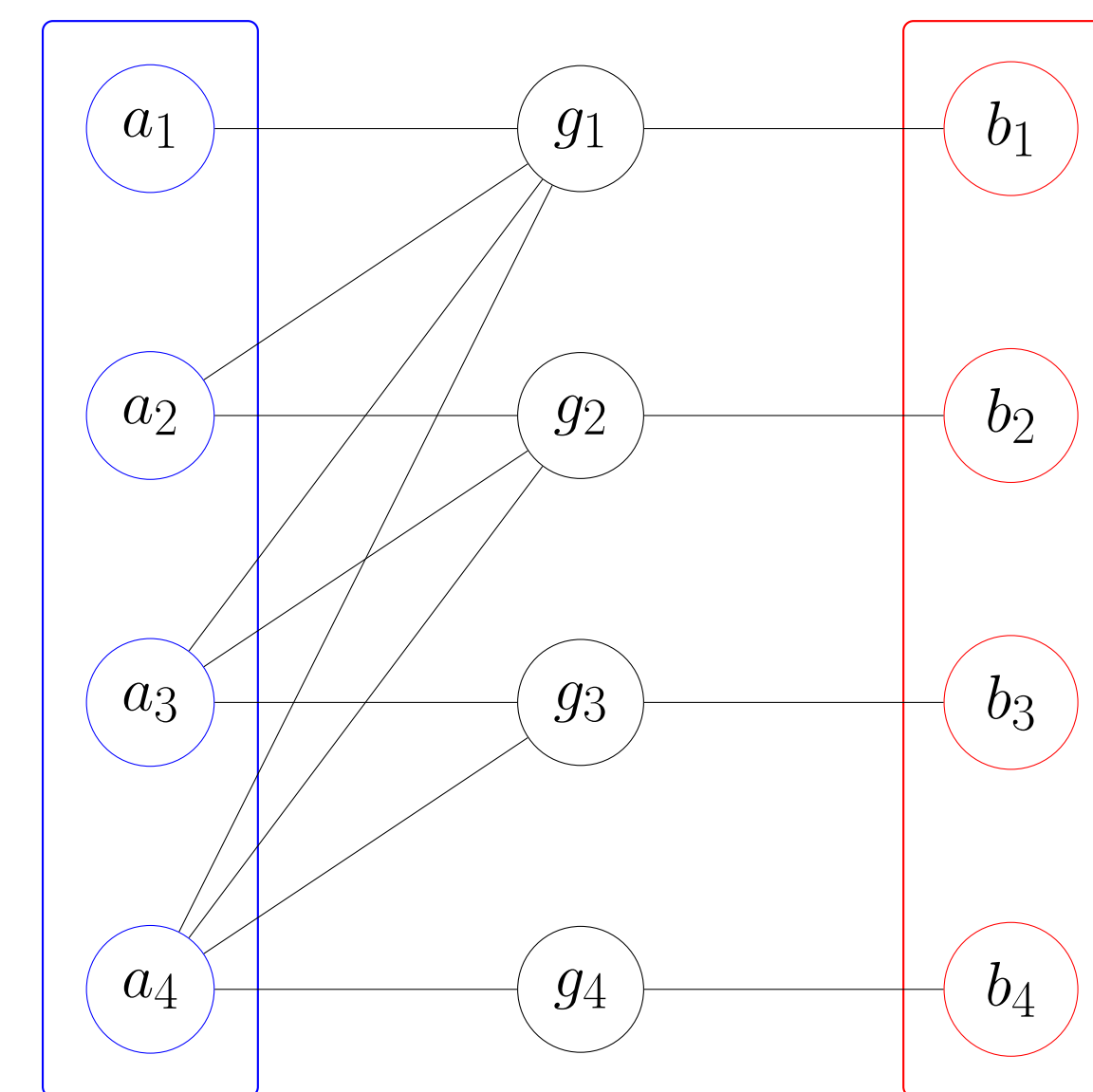
- A matching is α -CEF if for all classes $i, j \in [k]$, $v_i(A_i) \geq \alpha \cdot v_i^*(A_j)$ where $v_i^*(\cdot)$ denote class i 's "optimistic" value.
- Let μ_i be the value for each class when asked to maximize the worst partition it can create when optimally allocating items $\mu_i = \max_{M \in \mathcal{M}} \min_{j \in [k]} v_i^*(X_j)$ where \mathcal{M} is the set of all indivisible matchings. A matching is α -CMMS if for every class $i \in [k]$, $v_i(A_i) \geq \alpha \cdot \mu_i$.

Efficiency Definition: Maximize the number of matched agents (so-called "utilitarian social welfare")

Impossibility Results

No better than $1 - 1/e$ -CEF:

Consider $k = 2$ classes of n and n arriving items:



Price of Fairness:

- For some α , choose integers p, q such that $|\alpha - \frac{p}{q}| < \epsilon$.
- k classes: classes 1 to $k - 1$ have q agents, the k -th class has $q(k - 1)$.

Adversarial input sequence:

- **Phase One:**
 - $p(k - 1) + q$ items arrive, each item is liked by every agent (edges to all agents).
- **Phase Two:**
 - $k - 1$ groups of items arrive sequentially.
 - **Group i :** Contains q items, each item is liked only by agents in **Class i** .

Algorithm Overview

Algorithm 1 Random

```

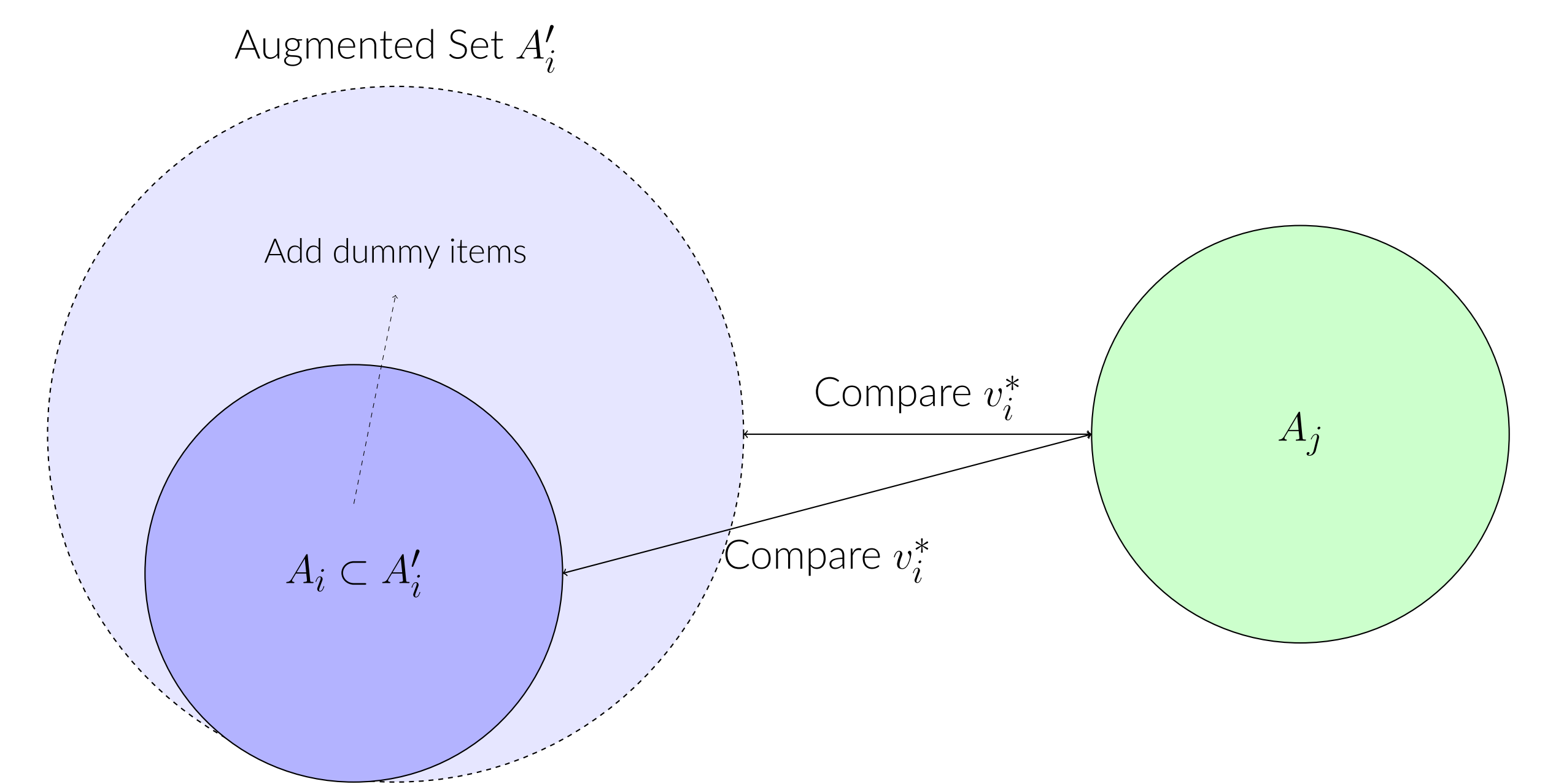
1: for  $o \in M$  do
2:    $S_o \leftarrow \emptyset$ 
3:   for  $i \in [k]$  do
4:     if  $\exists a \in N_i$  s.t.  $(a, o) \in E$  and  $x_a = 0$  then
5:        $S_o \leftarrow S_o \cup \{i\}$ 
6:     end if
7:   end for
8:   Pick an  $i \in S_o$  uniformly at random
9:   Pick an  $a \in N_i$  with  $(a, o) \in E$  and  $x_a = 0$  uniformly at random
10:  Set  $x_{a,o} = 1$ 
11: end for

```

Key Takeaways

- We provide the first non-wasteful algorithm that simultaneously obtains approximate class fairness guarantees in expectation, resolving a major open problem posed by [2] in an affirmative manner.
- We demonstrate that the expected guarantees are almost tight, but conjecture that the optimal competitive ratio for randomized algorithms is $1 - 1/e$.
- We provide a strengthened upper bound for the *divisible* matching setting, further resolving an open problem left by [2].
- Defined the "price of fairness," the necessary trade-off between an optimal and a fair matching.

Proof Idea



1. $v_i^*(A'_i) \leq 2 \cdot v_i(A_i)$
2. $\mathbf{E}[v_i^*(A'_i)] \geq \mathbf{E}[v_i^*(A_j)]$
3. $\Rightarrow \mathbf{E}[v_i(A_i)] \geq \frac{1}{2} \cdot \mathbf{E}[v_i^*(A_j)]$

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The full version of our paper can be found on arXiv with the QR-code above.

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