

**Tight FPT
Approximations for
k-Median and k-Means**

Jason Li

Joint work with

Vincent Cohen-Addad, Anupam Gupta, Amit Kumar, Euiwoong Lee

ICALP 2019

k-median problem

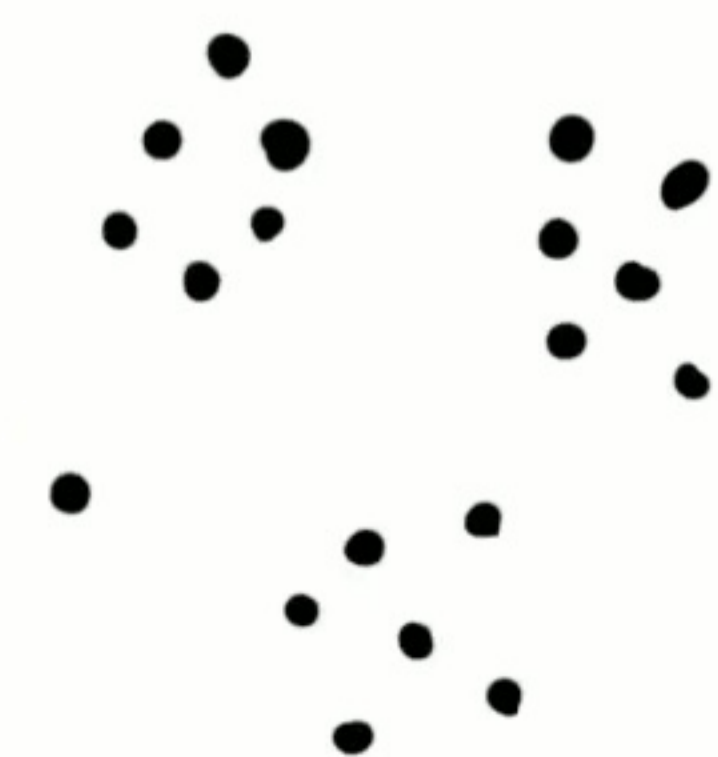
- Metric space (V, d)
- Clients $C \subseteq V$, Facilities $F \subseteq V$
- Find set F of k facilities minimizing

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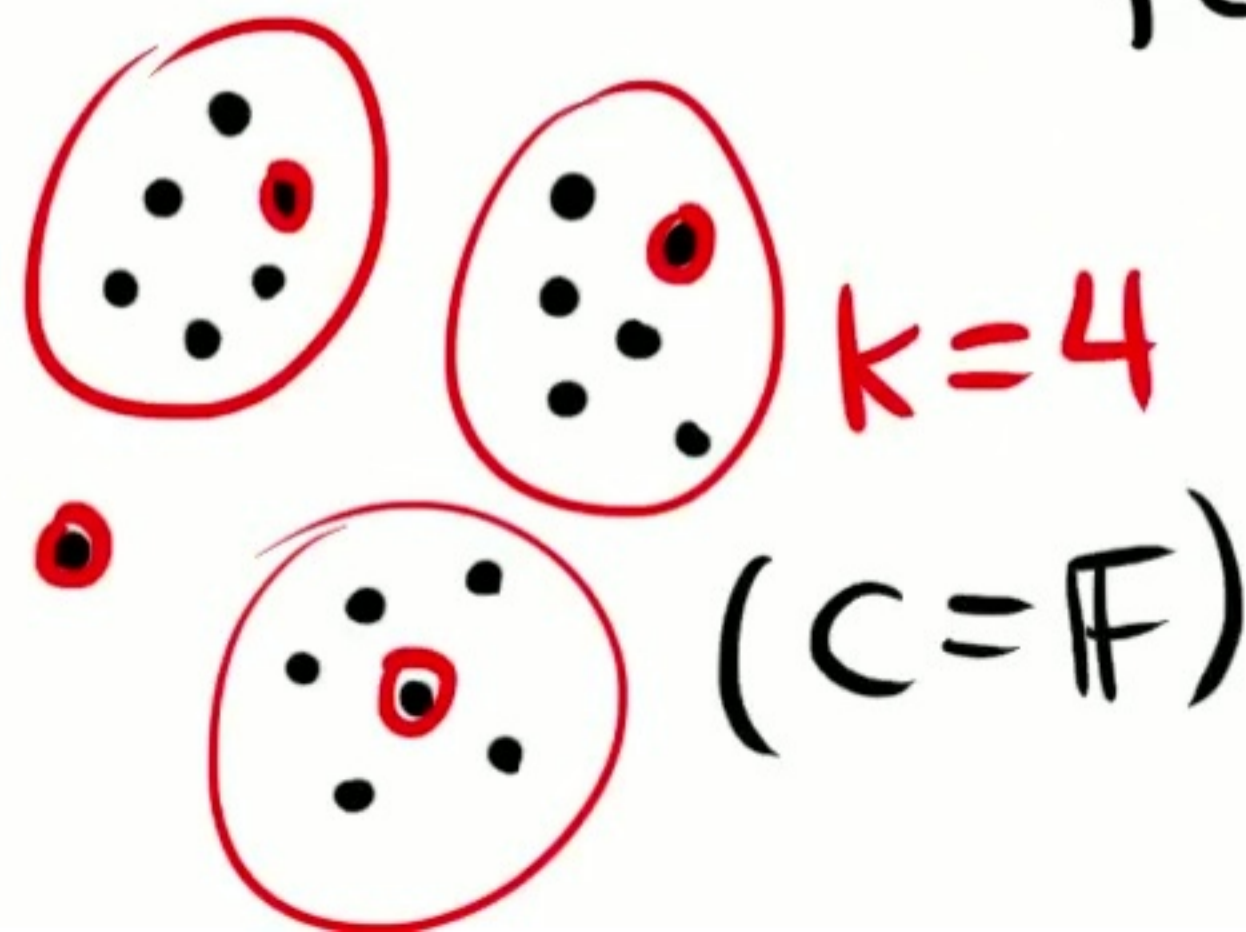
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- Clustering:  $k=4$
($C=F$)

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$$\min_{|F|=k} \sum_{v \in C} d(v, F)$$

- NP-hard (dominating set \leq k-median)

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Core-sets

Def[Core-set]: a subset $S \subseteq C$ with weights $w(v): v \in S$
s.t.

$\forall F$ set of k facilities:

$$\sum_{v \in S} w(v) \cdot d(v, F) \in (1 \pm \epsilon) \sum_{v \in C} d(v, F)$$

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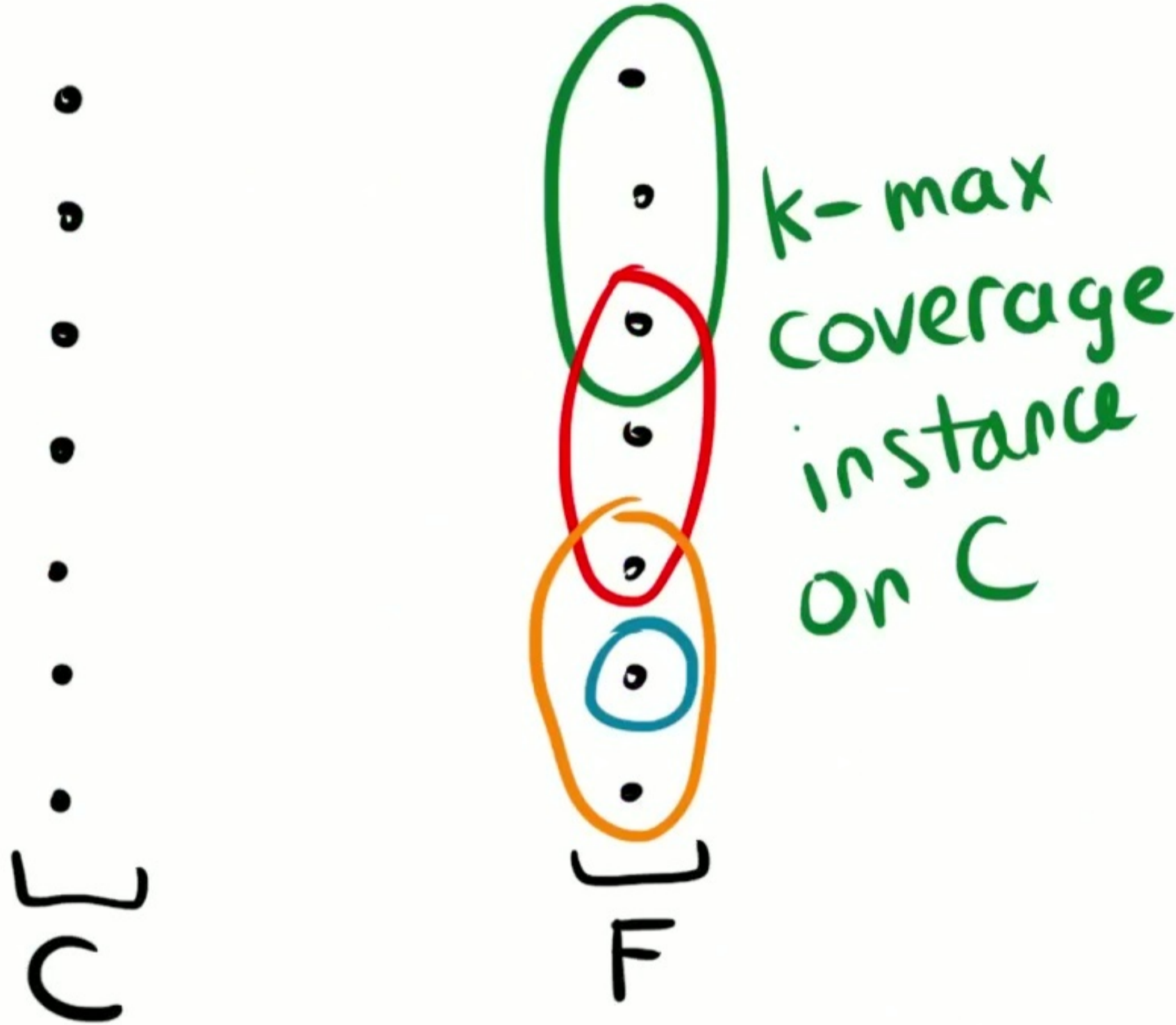
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Thm [Chen; Feldman-Langberg] \exists core-set size $\text{poly}(\epsilon^{-1} k \log n)$.
At loss of $(1+\epsilon)$, can assume $\text{poly}(\epsilon^{-1} k \log n)$ (weighted) clients

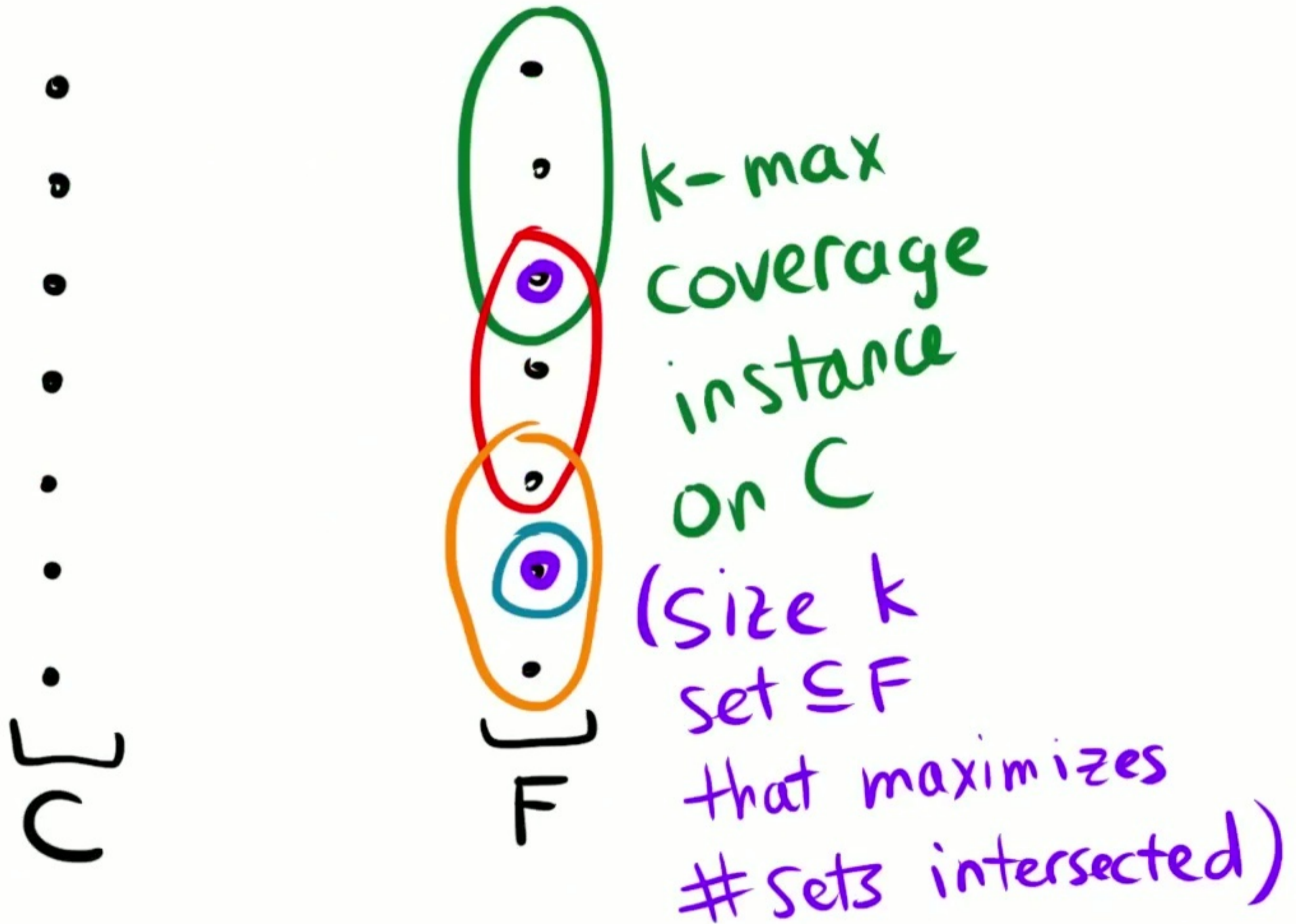
Insight from hardness: why $1+2/e$?



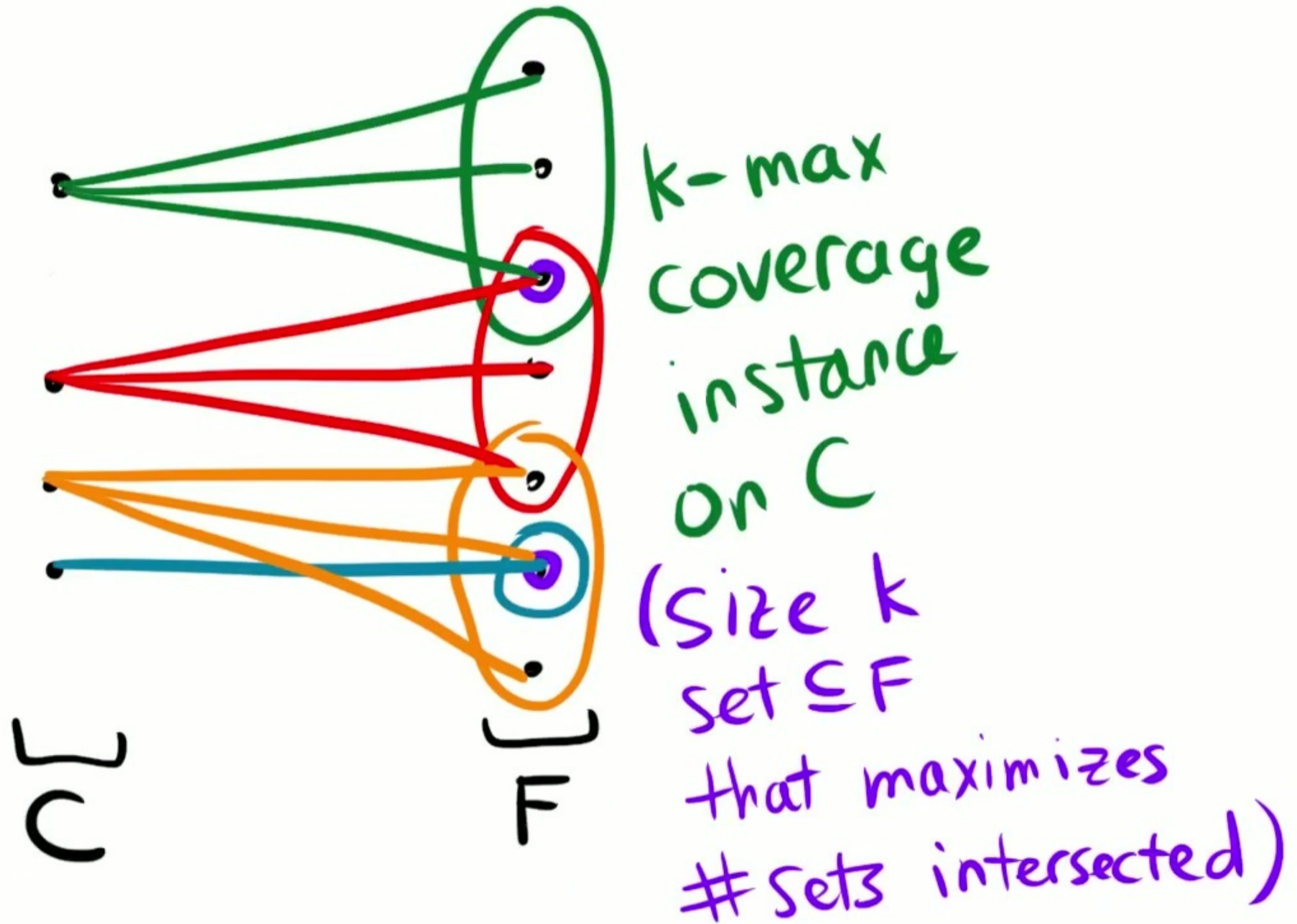
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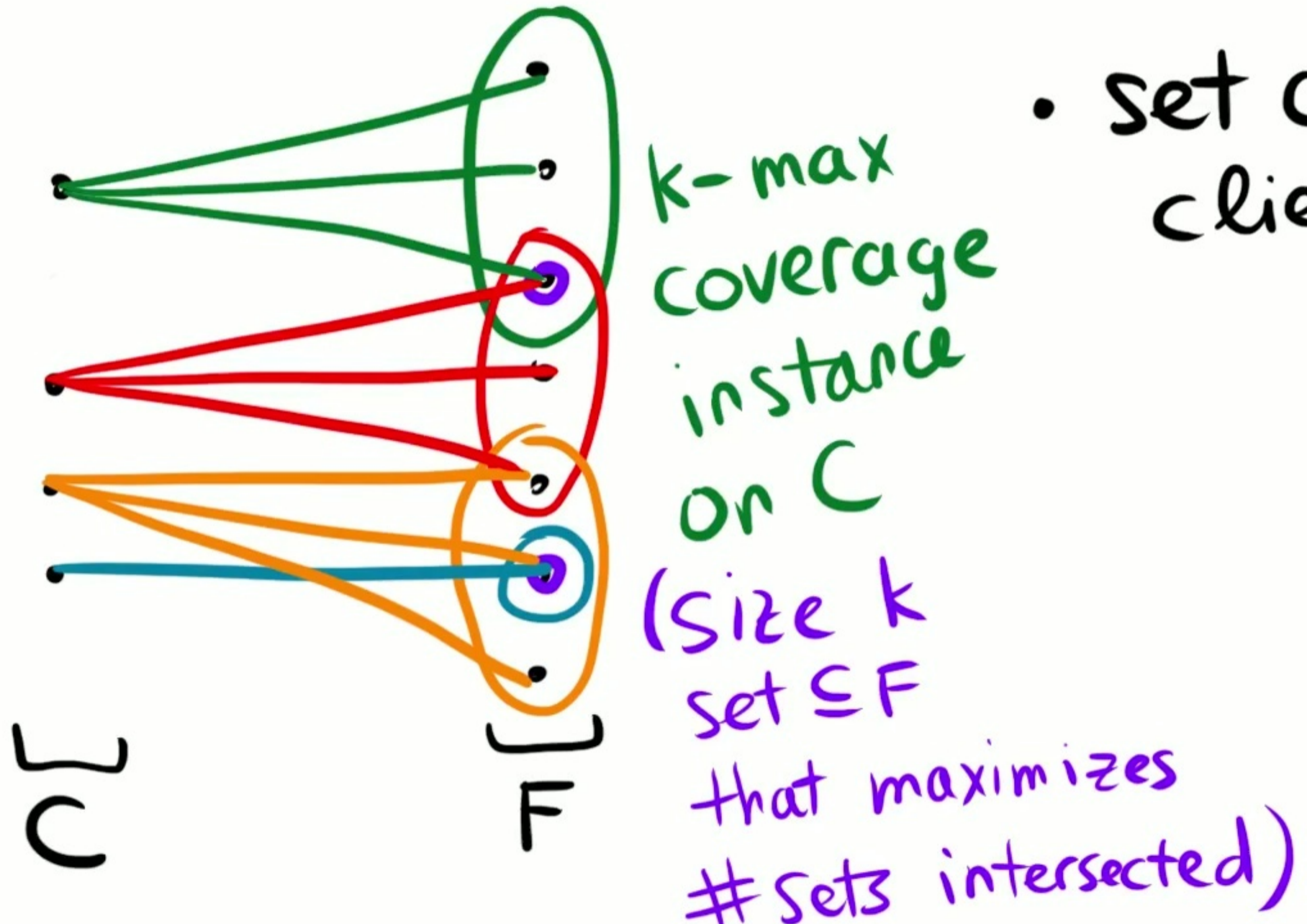
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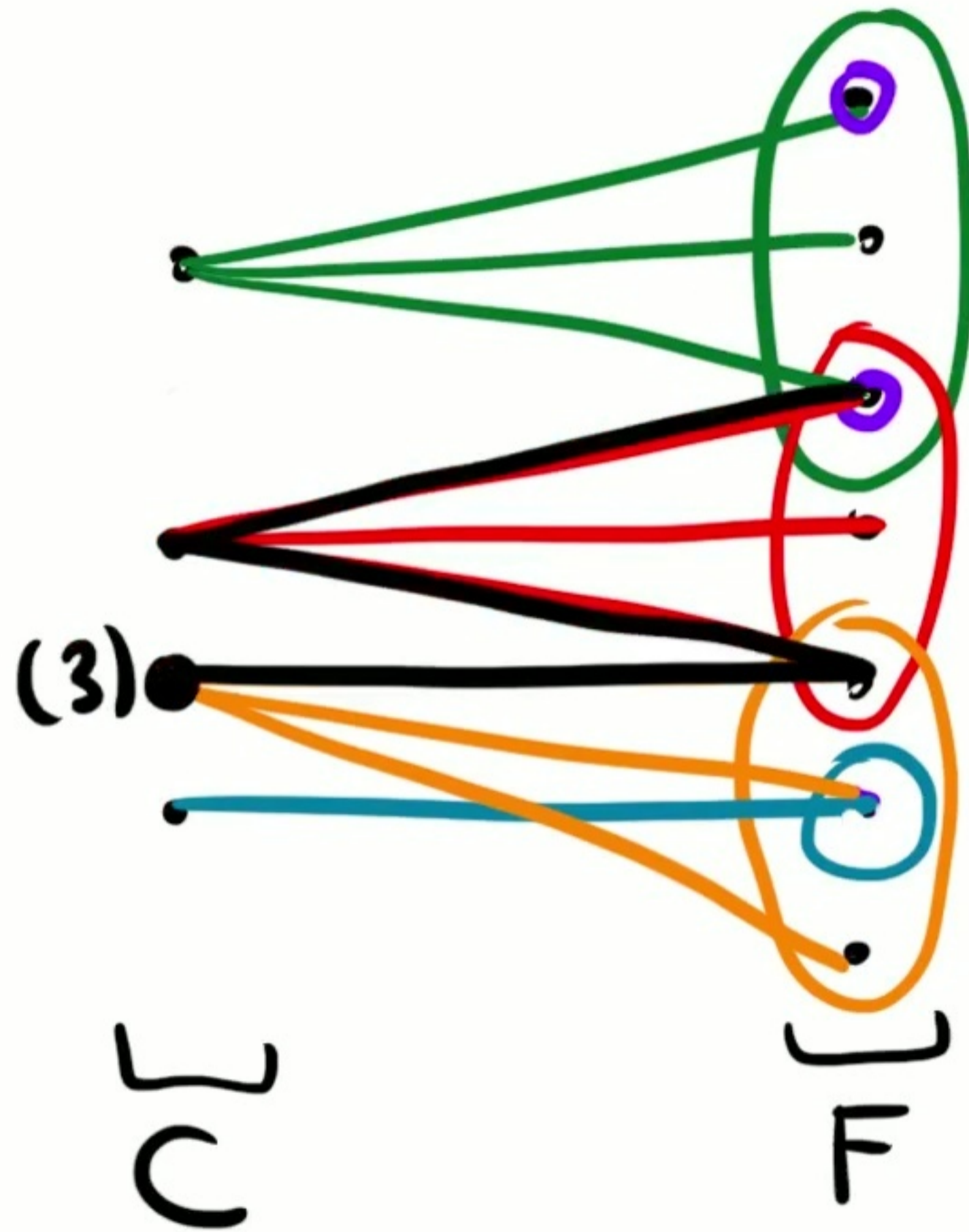


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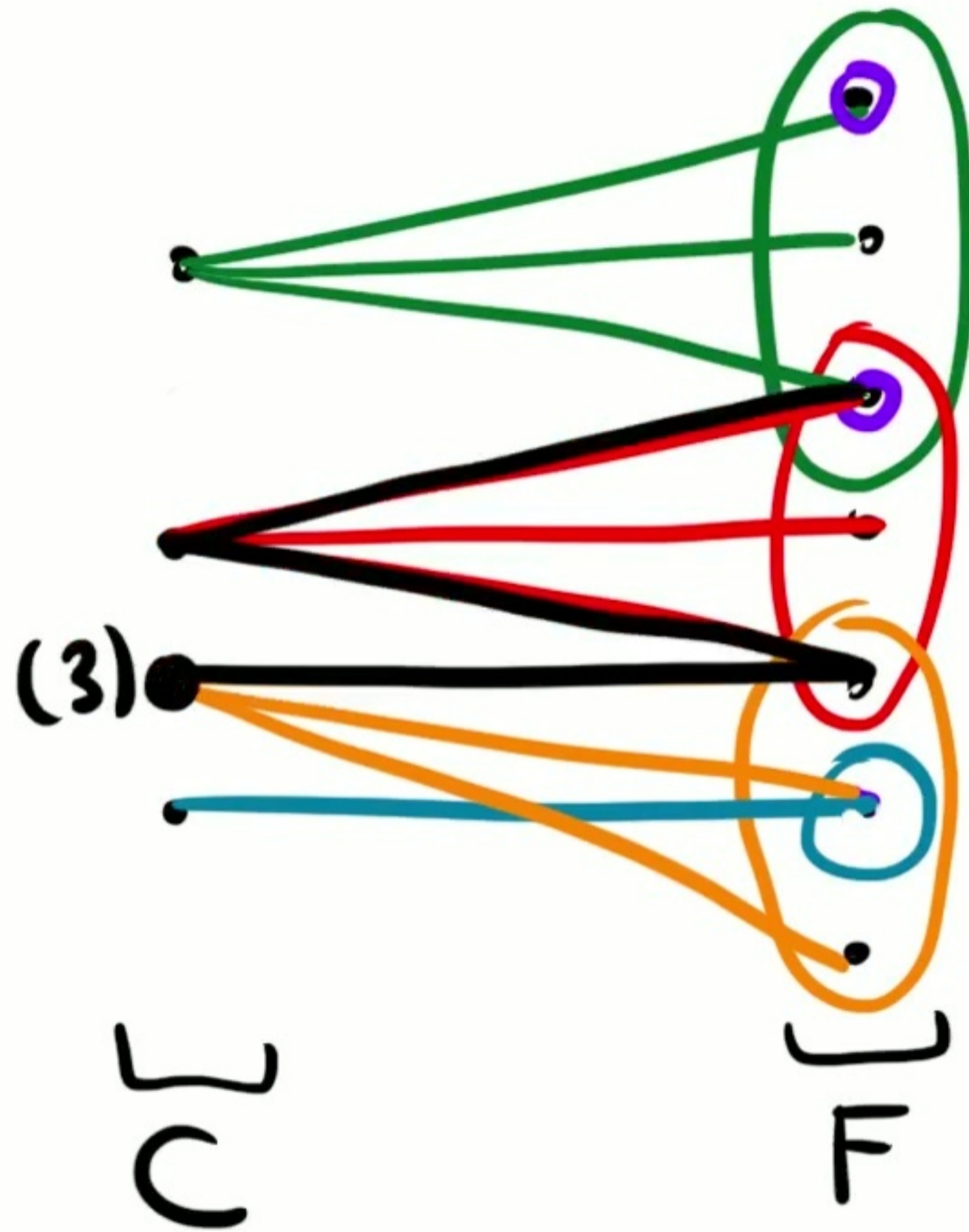


k-max
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instance
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(Size k
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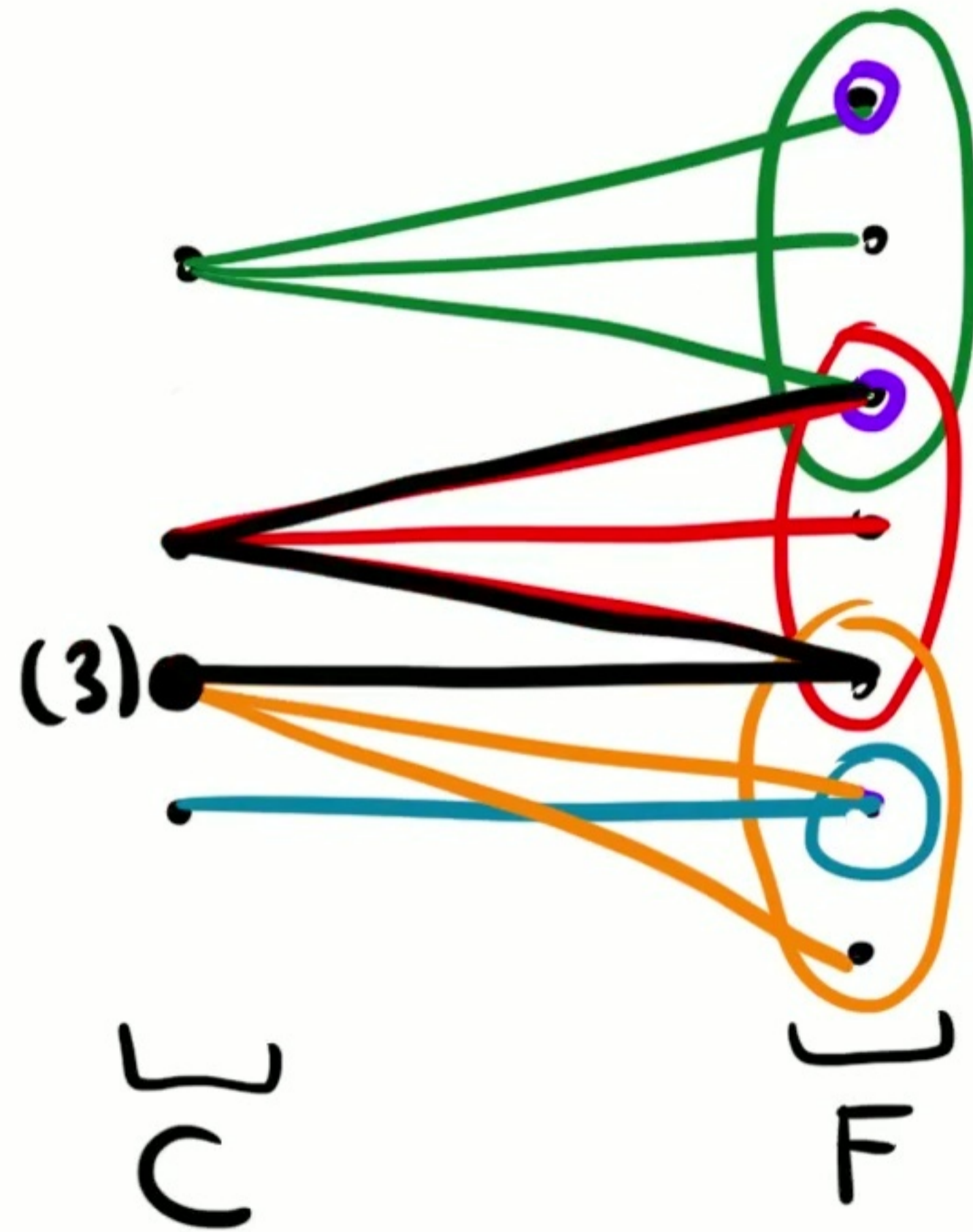


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\implies k-median $(1-1/e) \cdot 1 + \frac{1}{e} \cdot 3$

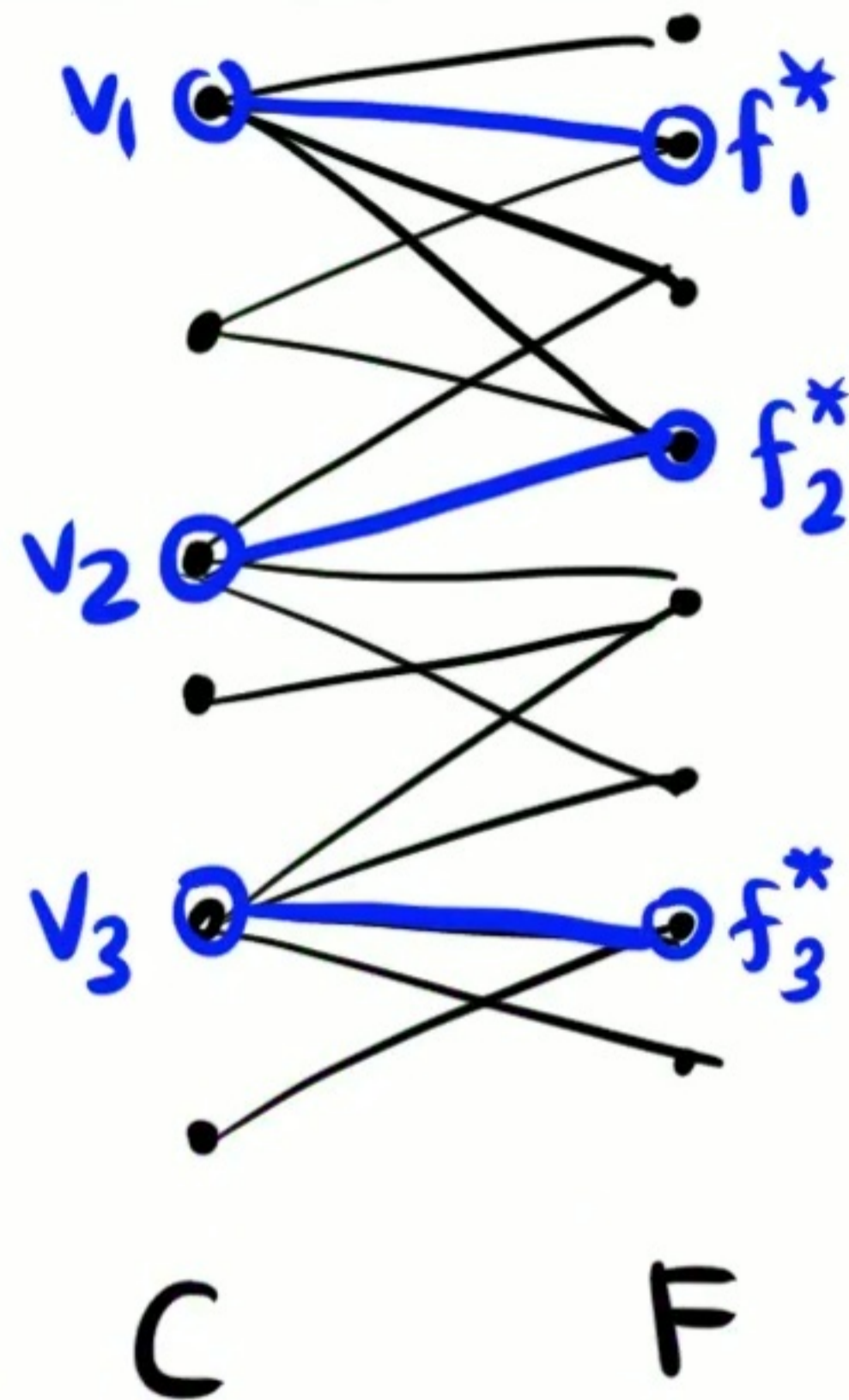
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Transformation to Algorithm

- Guess k centers $v_1 \dots v_k$

s.t. $f_i^* \sim v_i \forall i$

$[\text{poly}(\varepsilon^{-1} k \log n)]^k$ time (FPT!)



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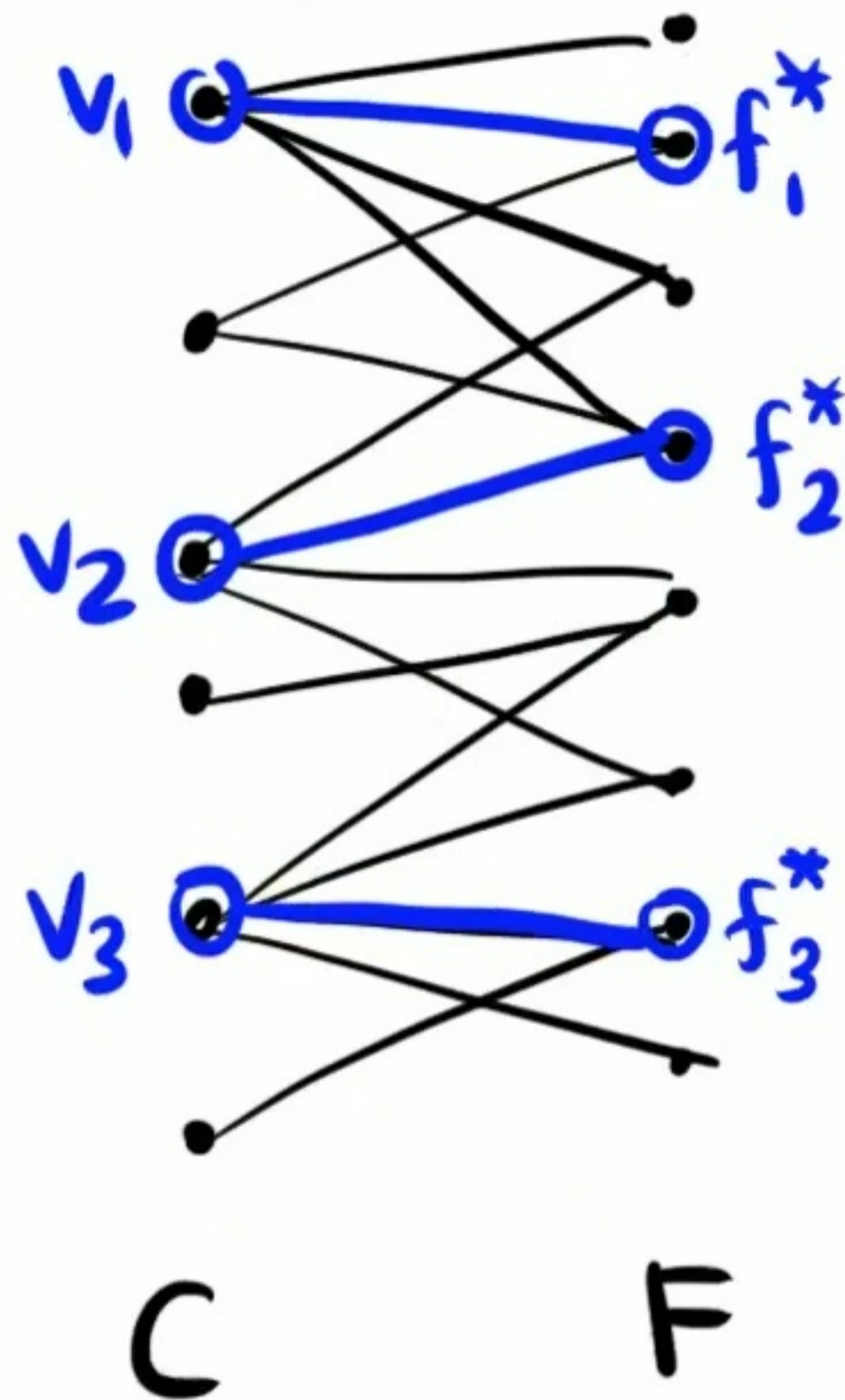
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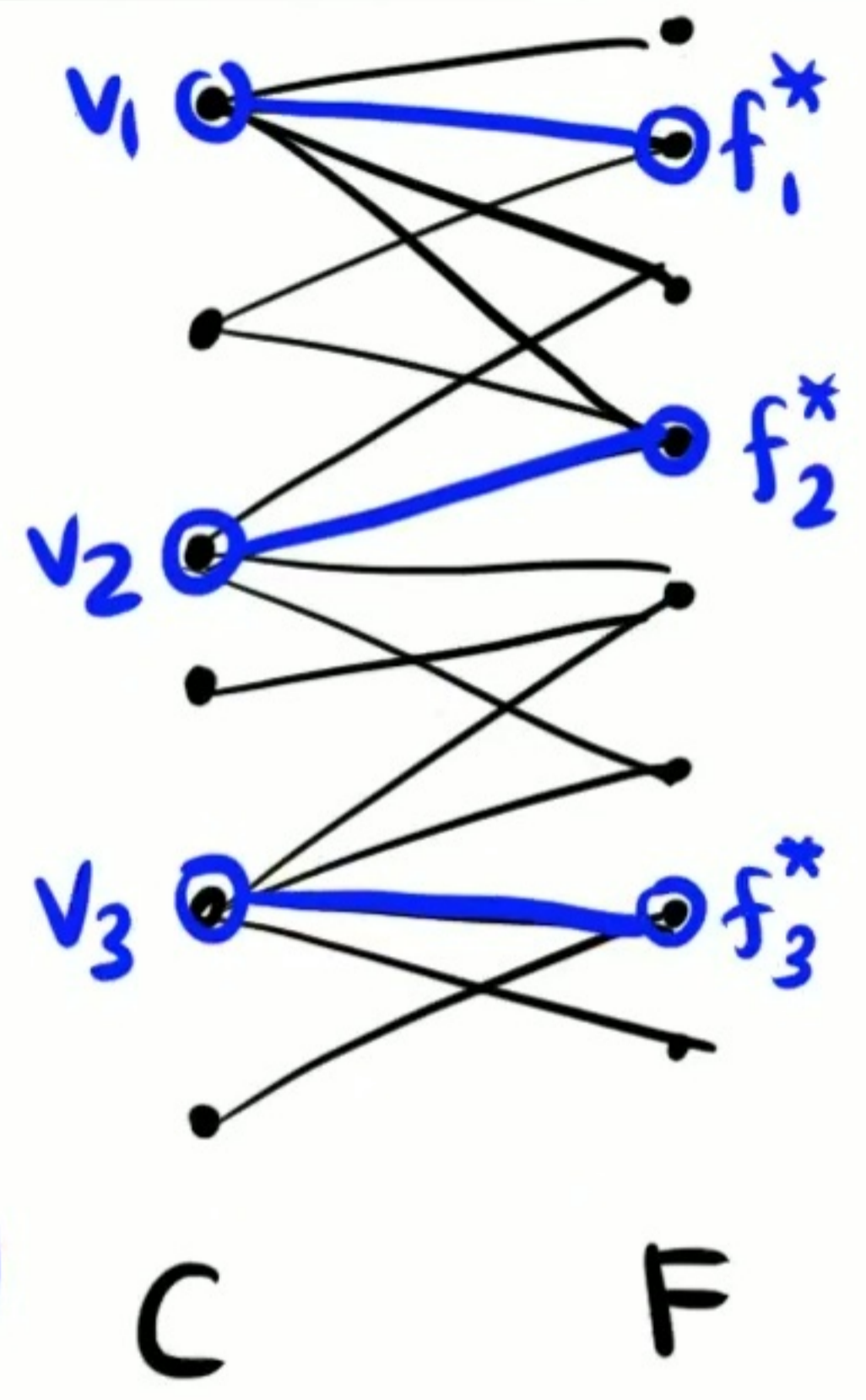


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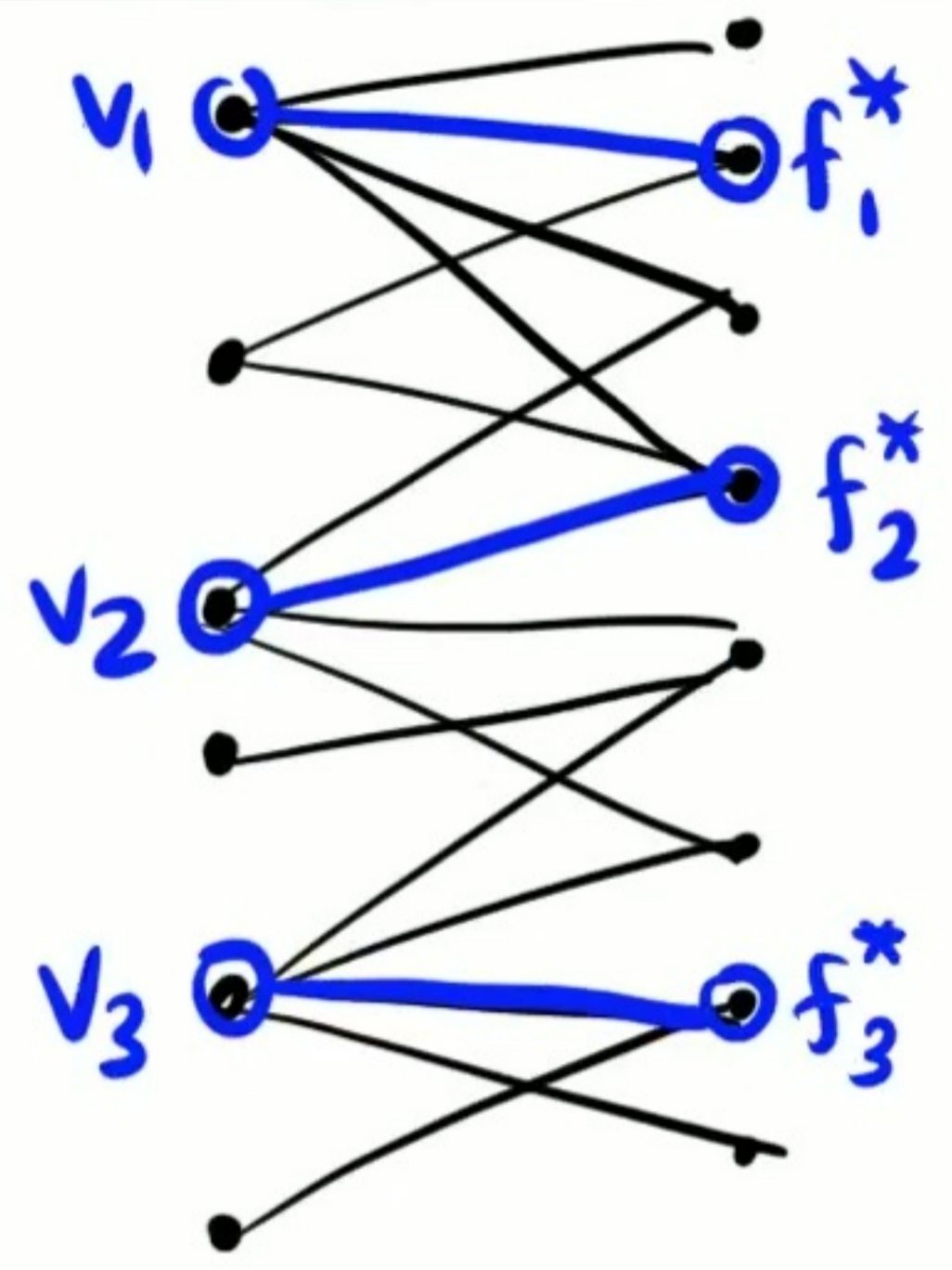
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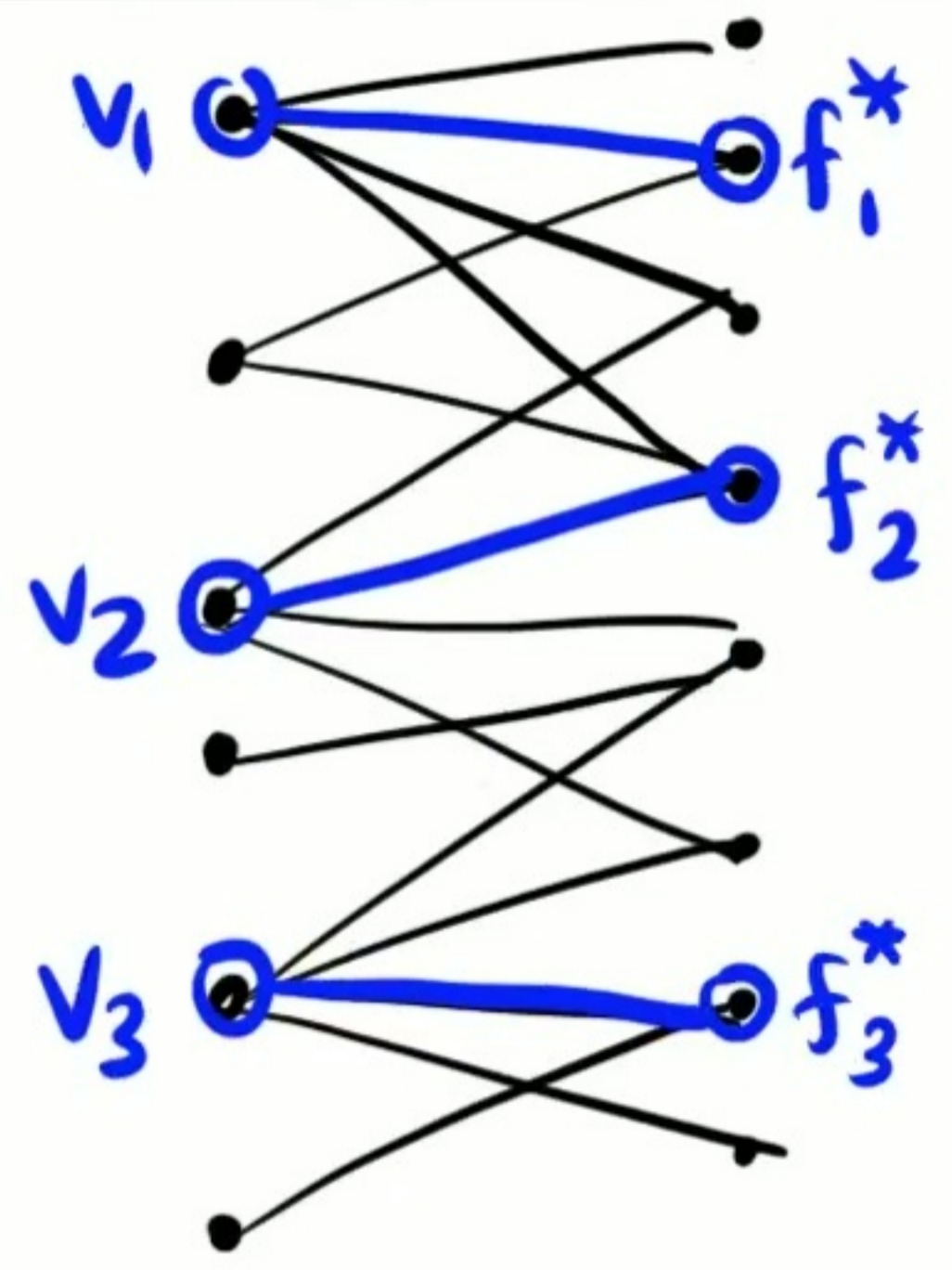
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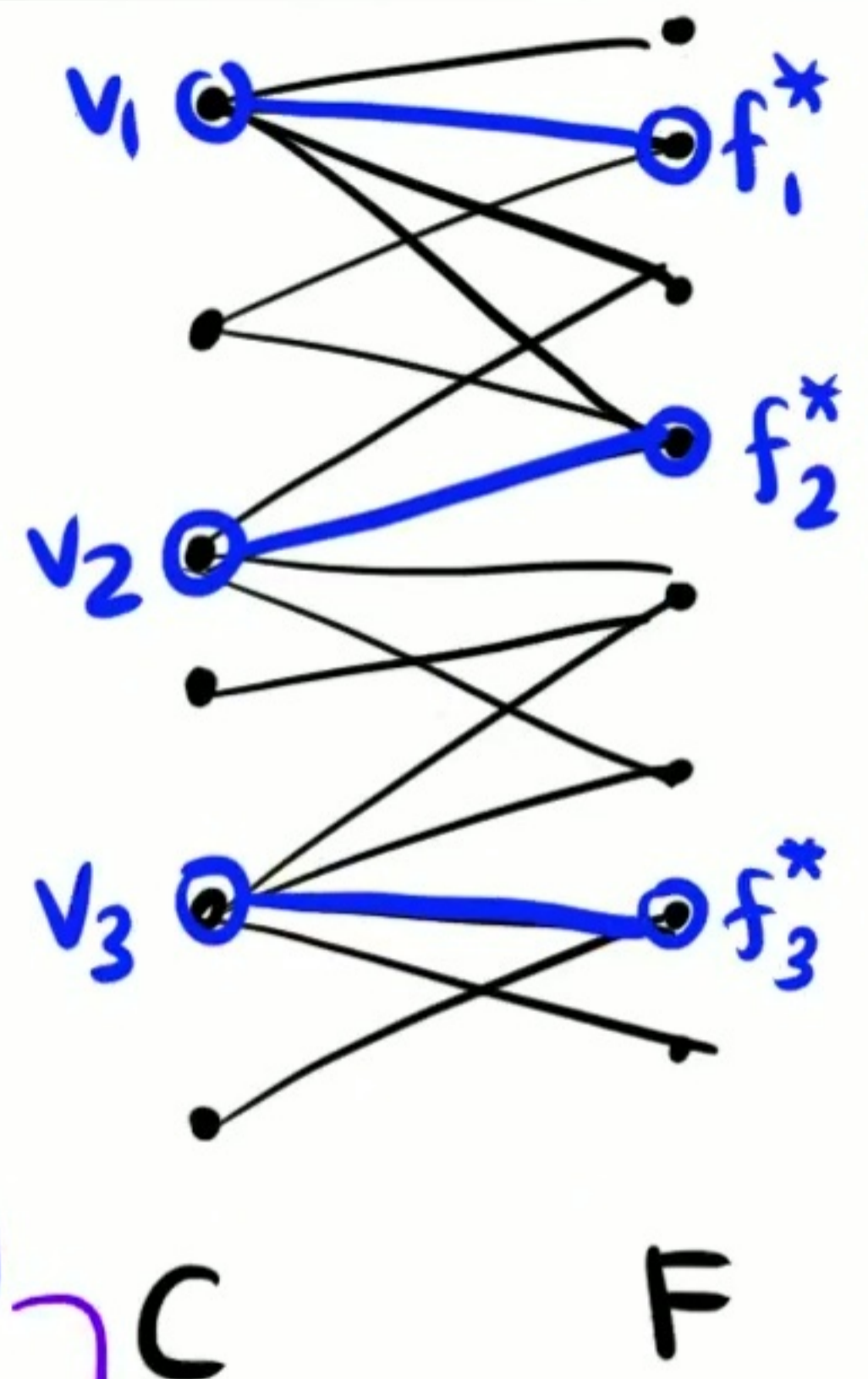
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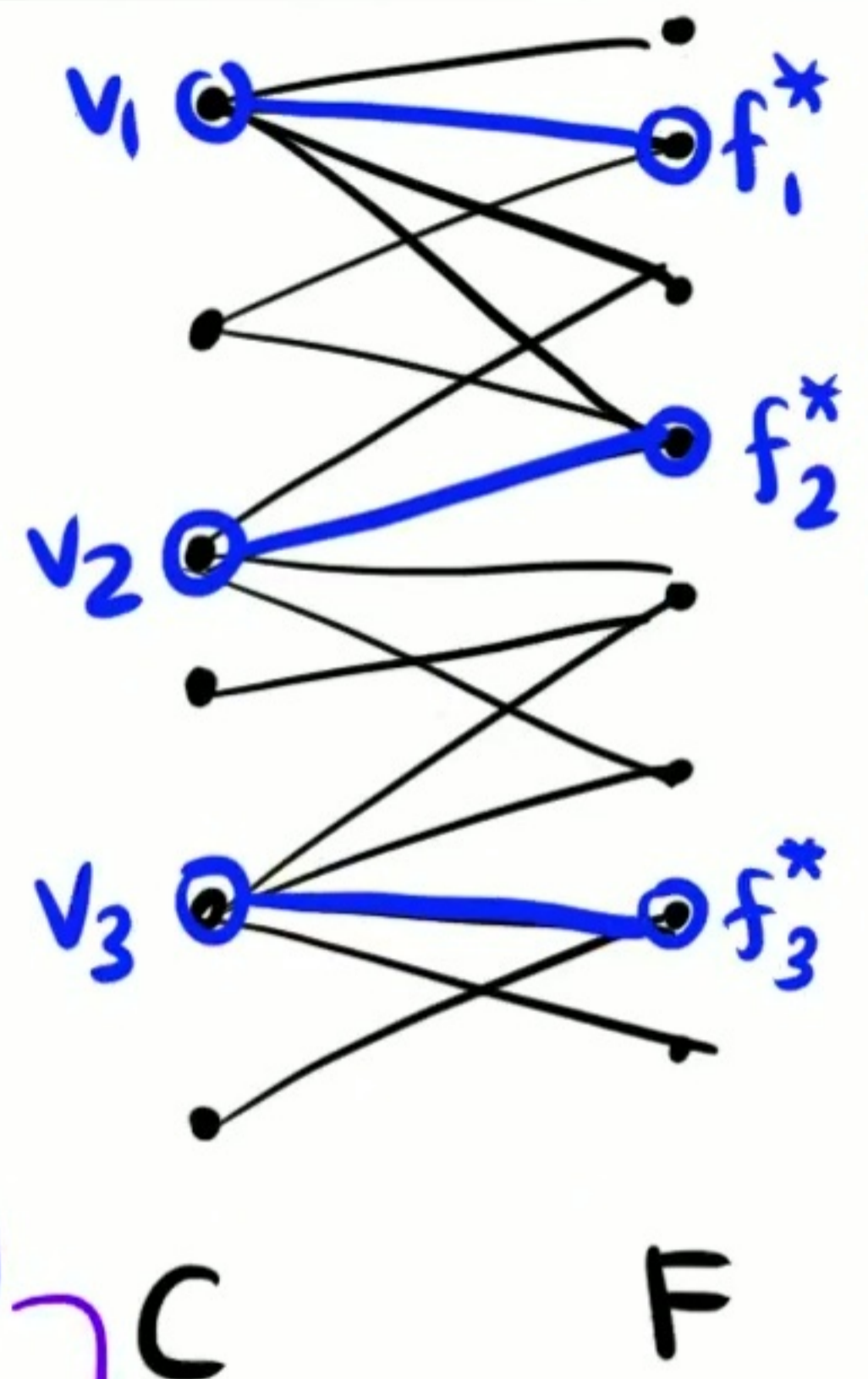
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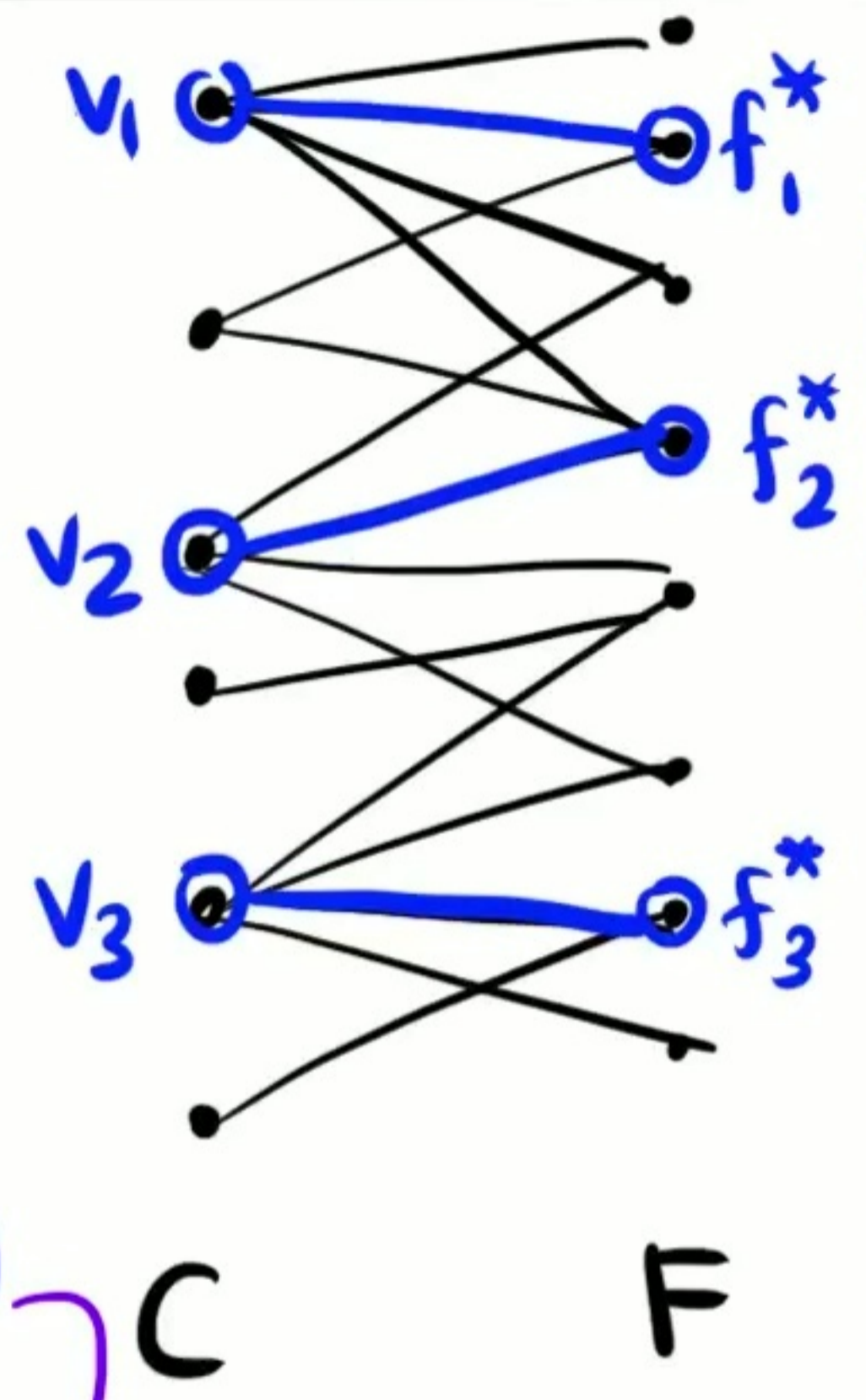
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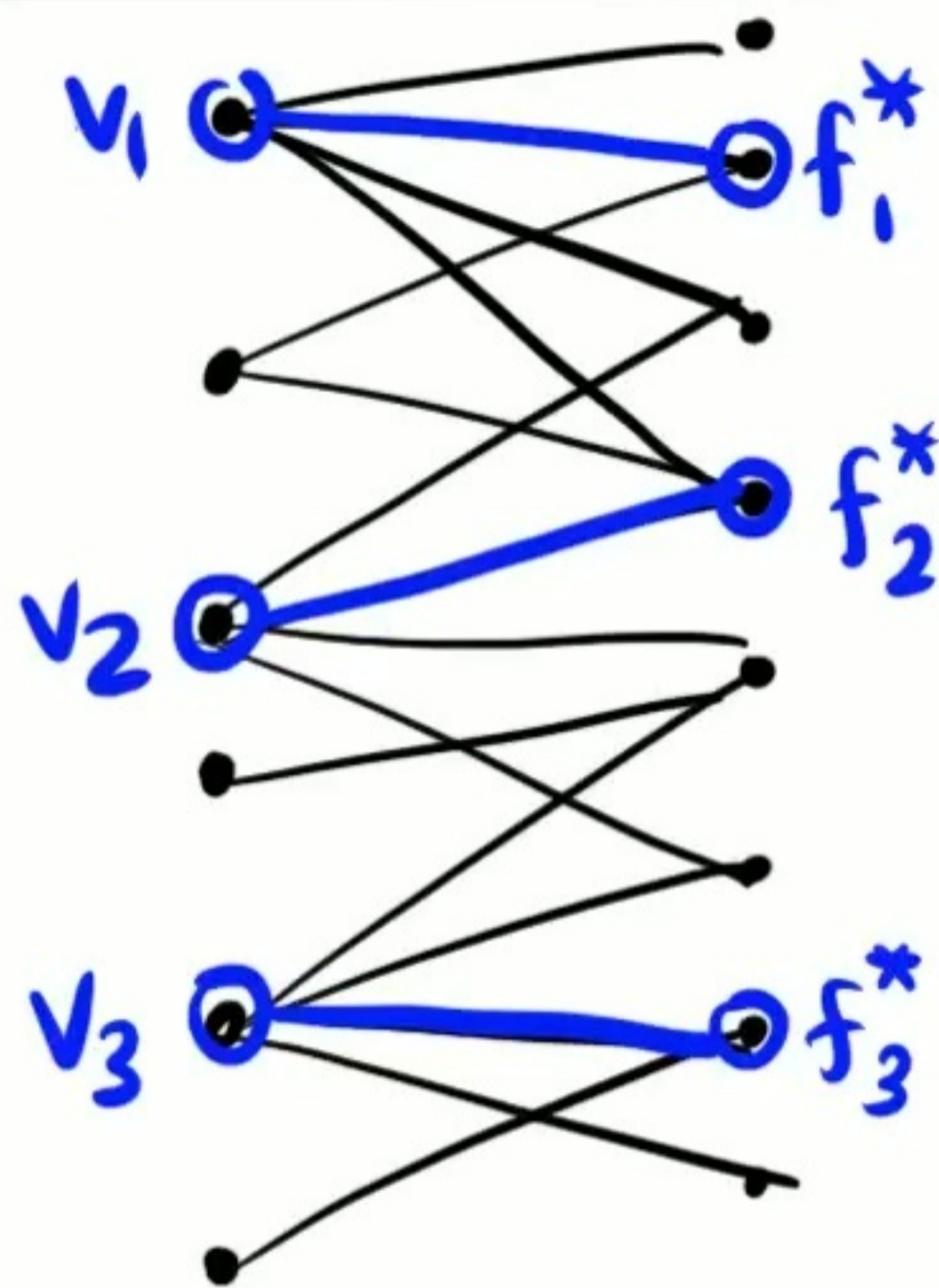
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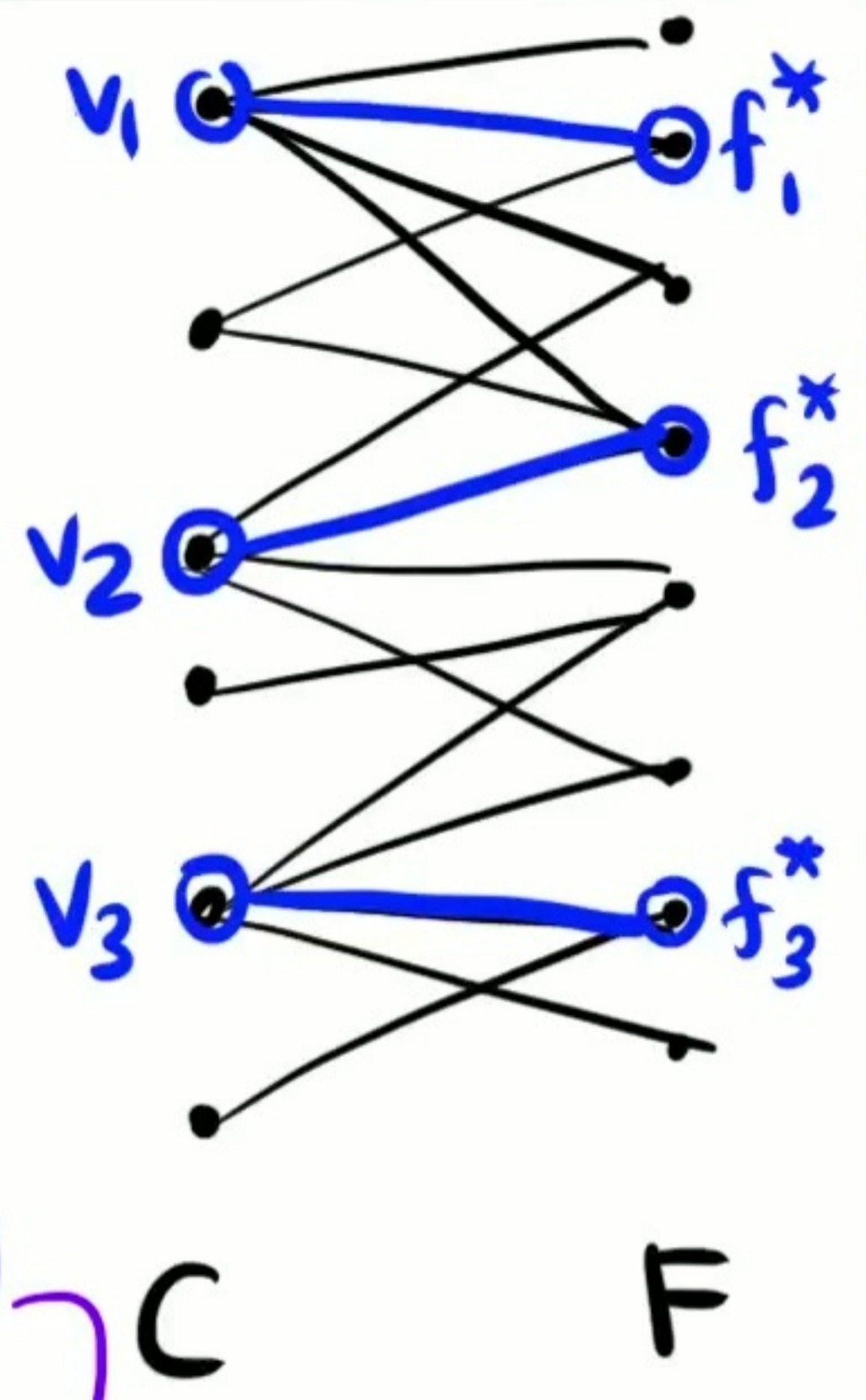
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No such guarantee in
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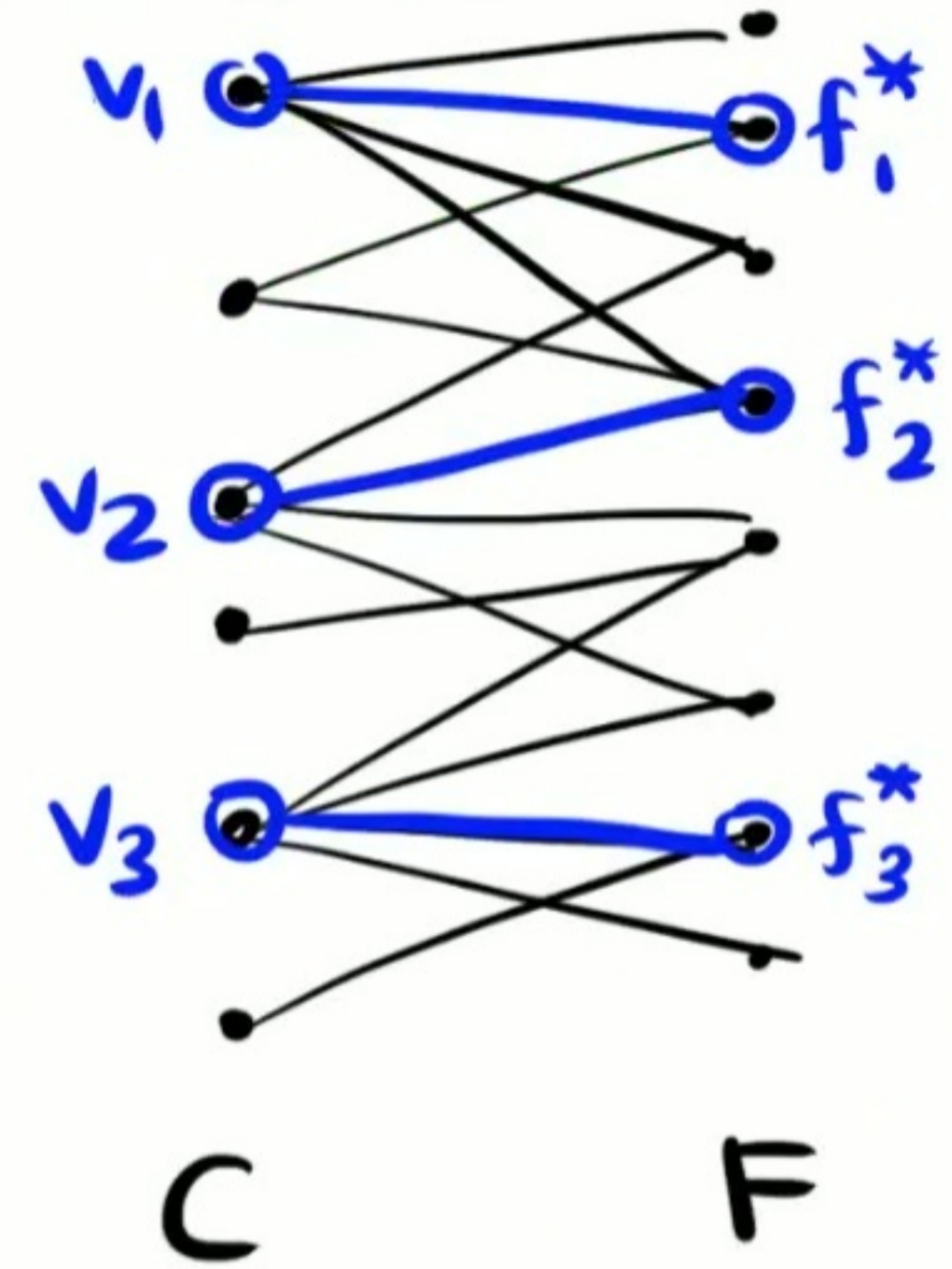
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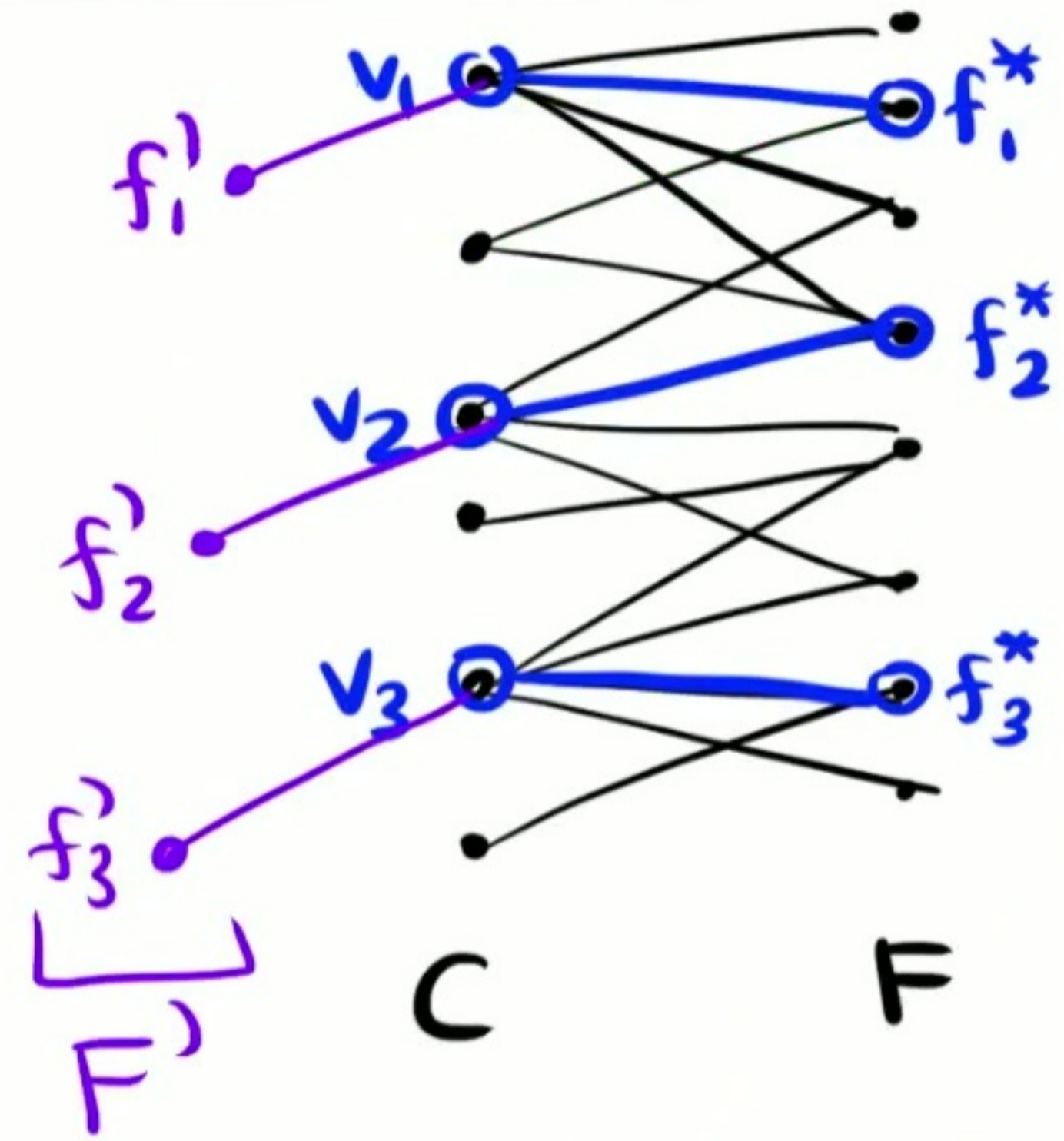
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→ 3-approx.
 How to get $(1+2/e)$? *don't know $d(v, F^*)$...*
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Optimization function



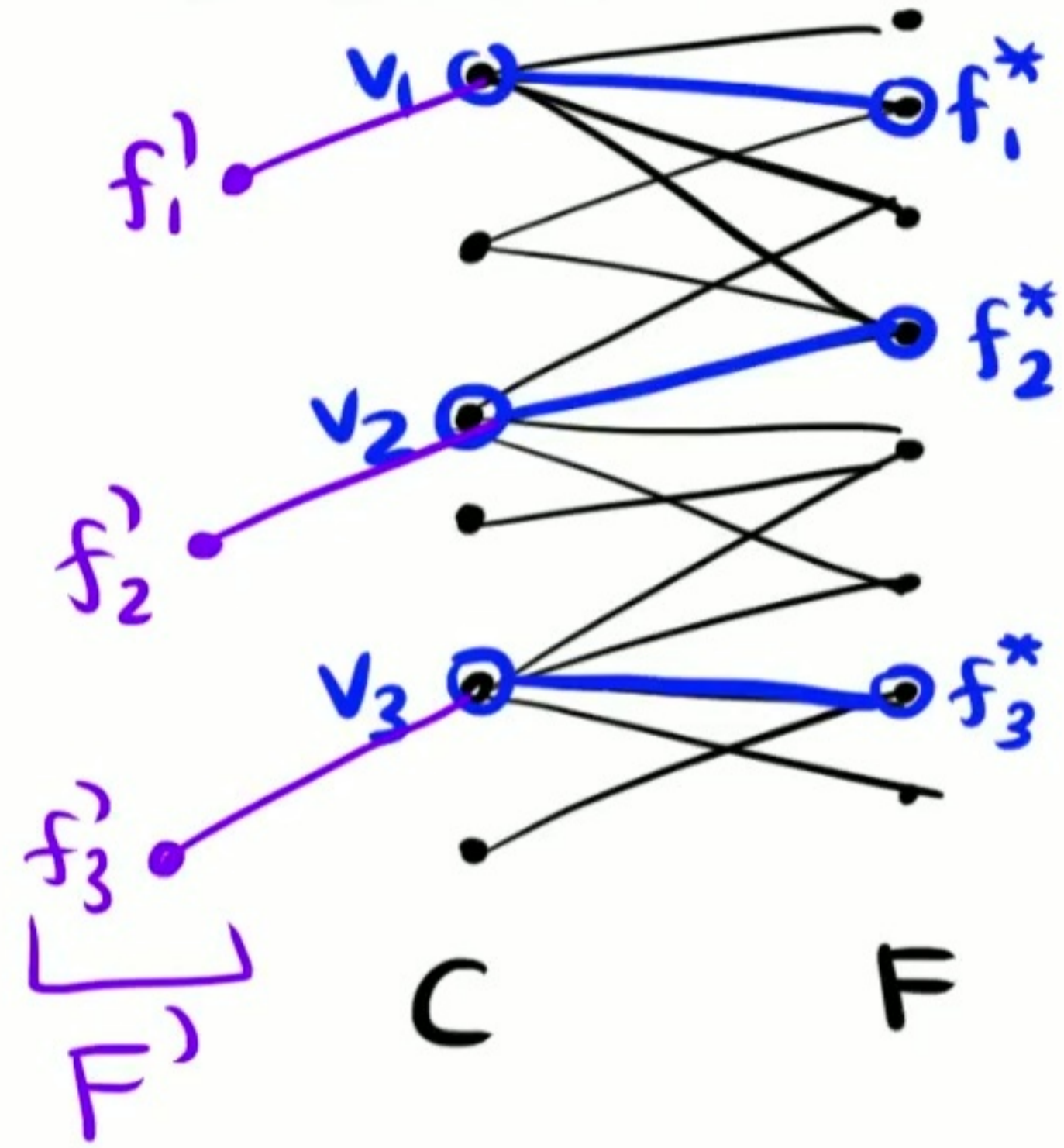
Optimization function



Optimization function

maximize
$$\text{impr}(S) := \sum_v w(v) (d(v, F') - d(v, F' \cup S))$$

over $S \subseteq F$, $S = \{f_1, \dots, f_k, f_i \in N(v_i)\}$



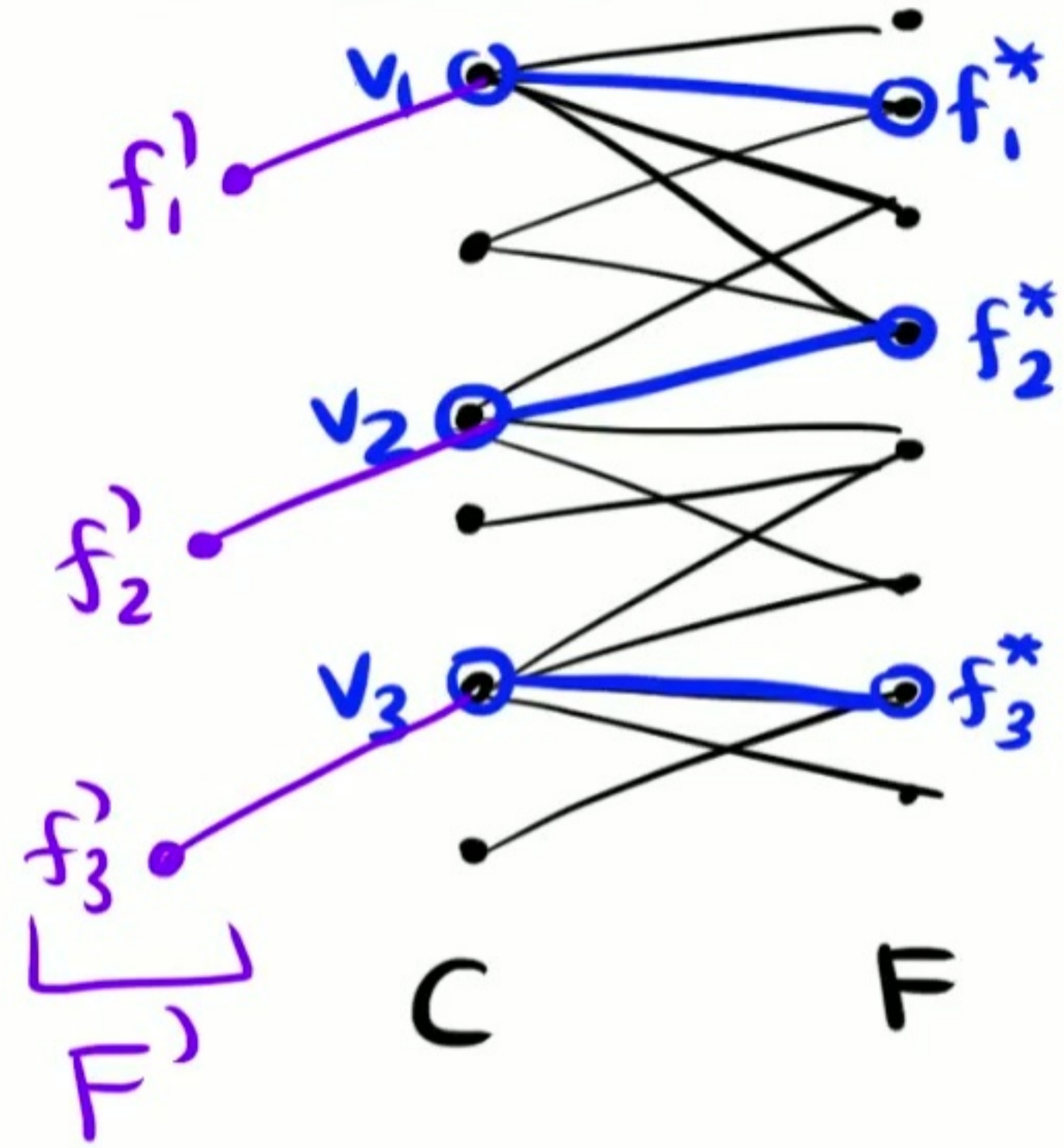
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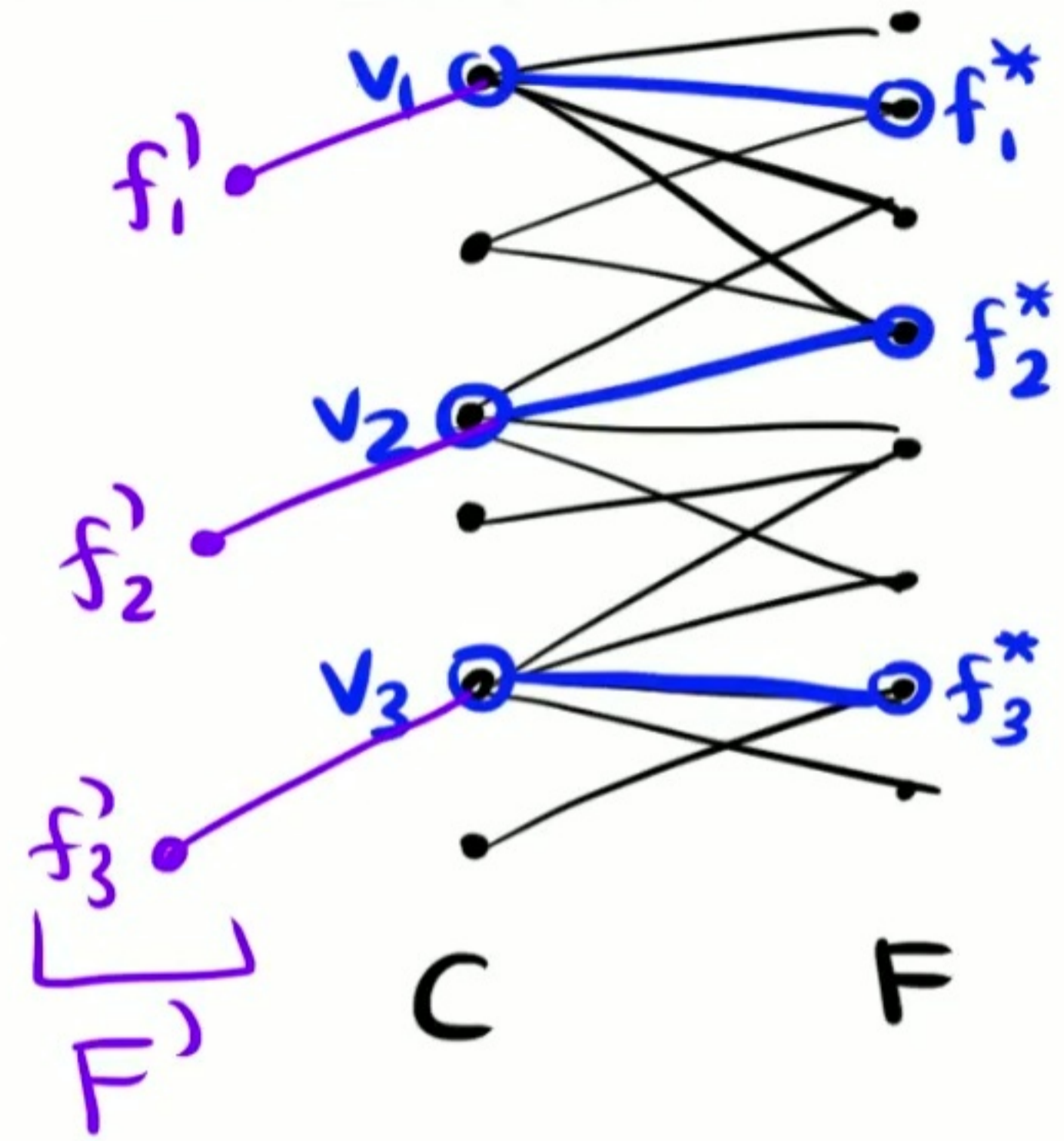
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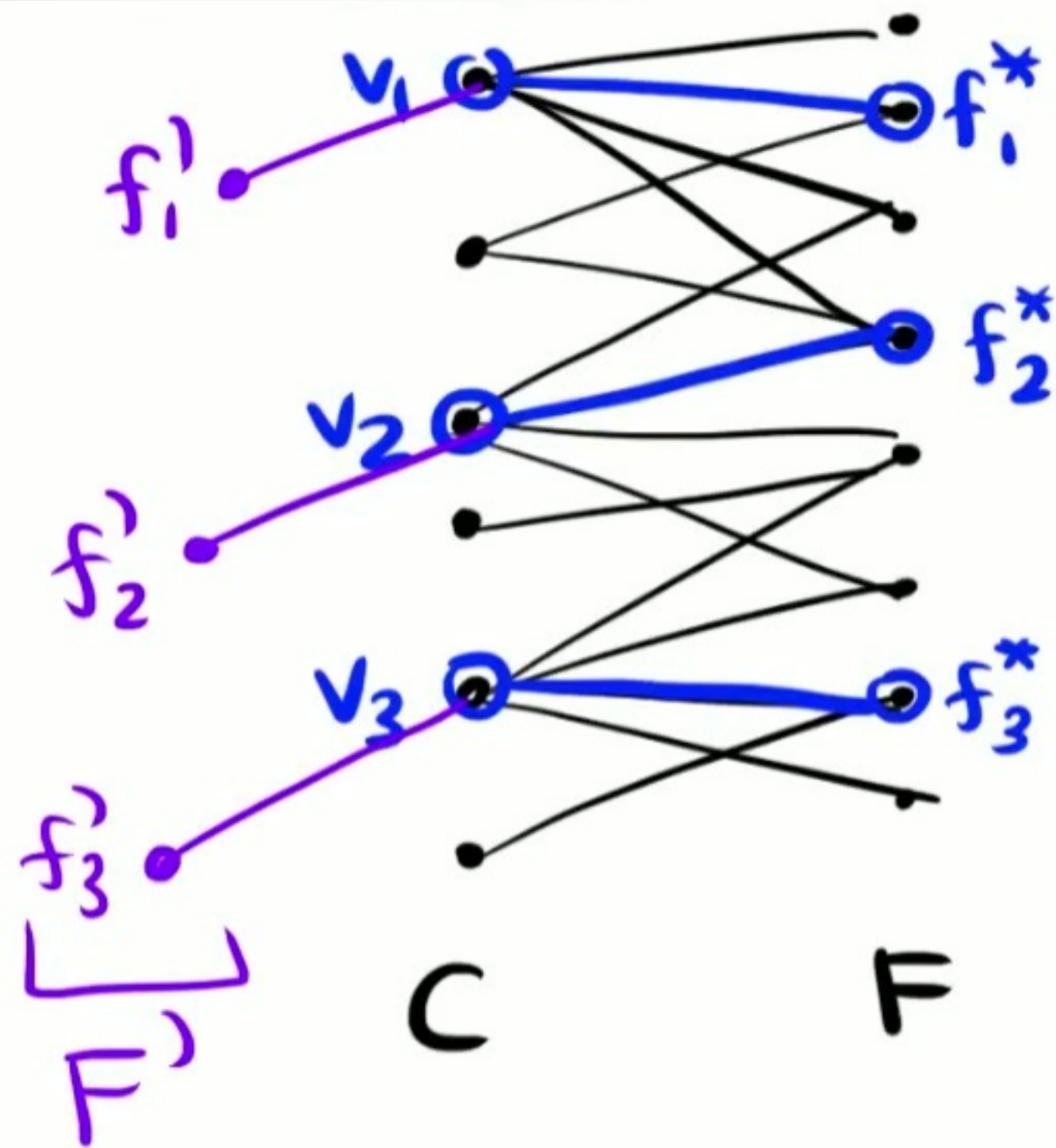
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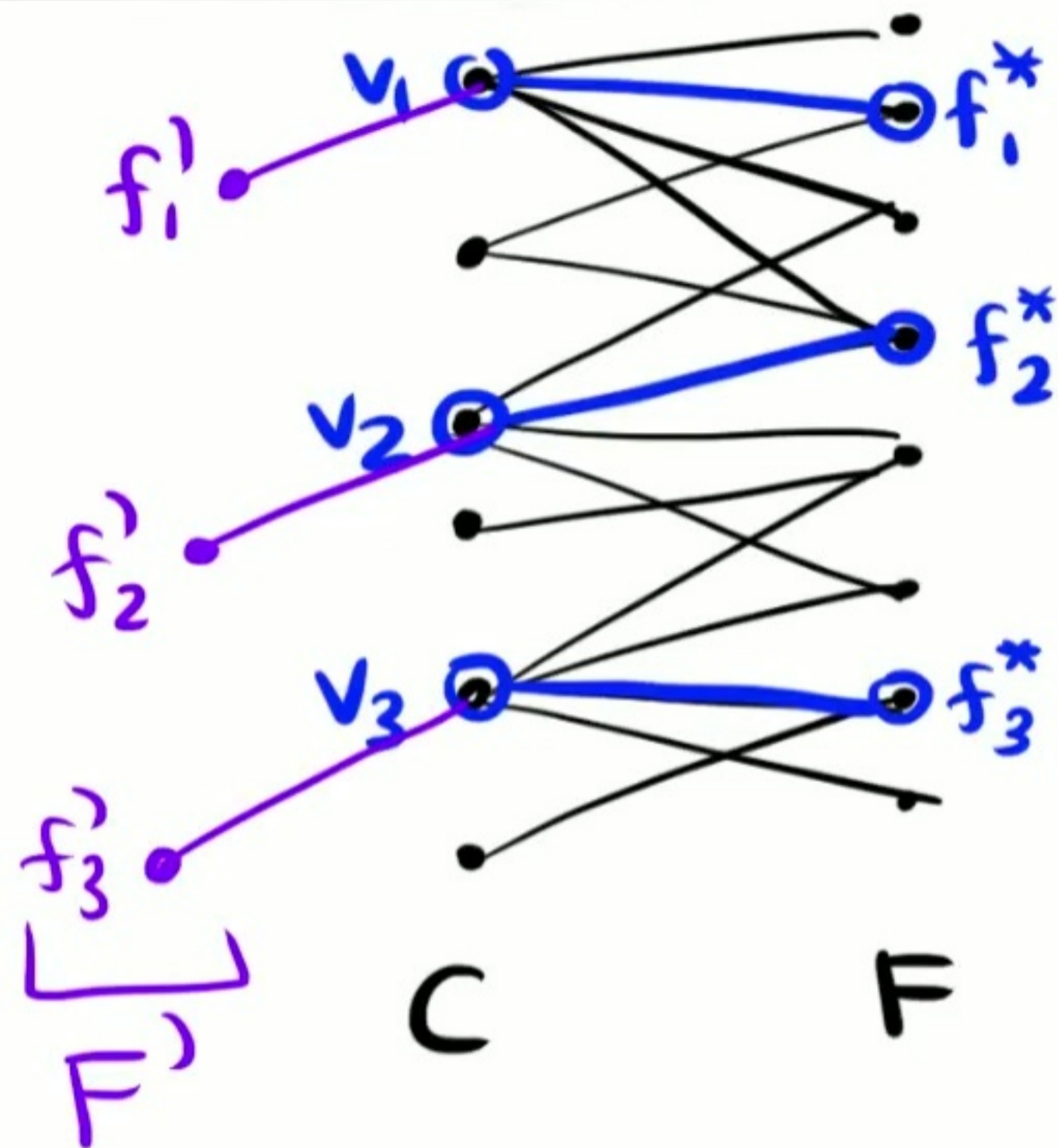
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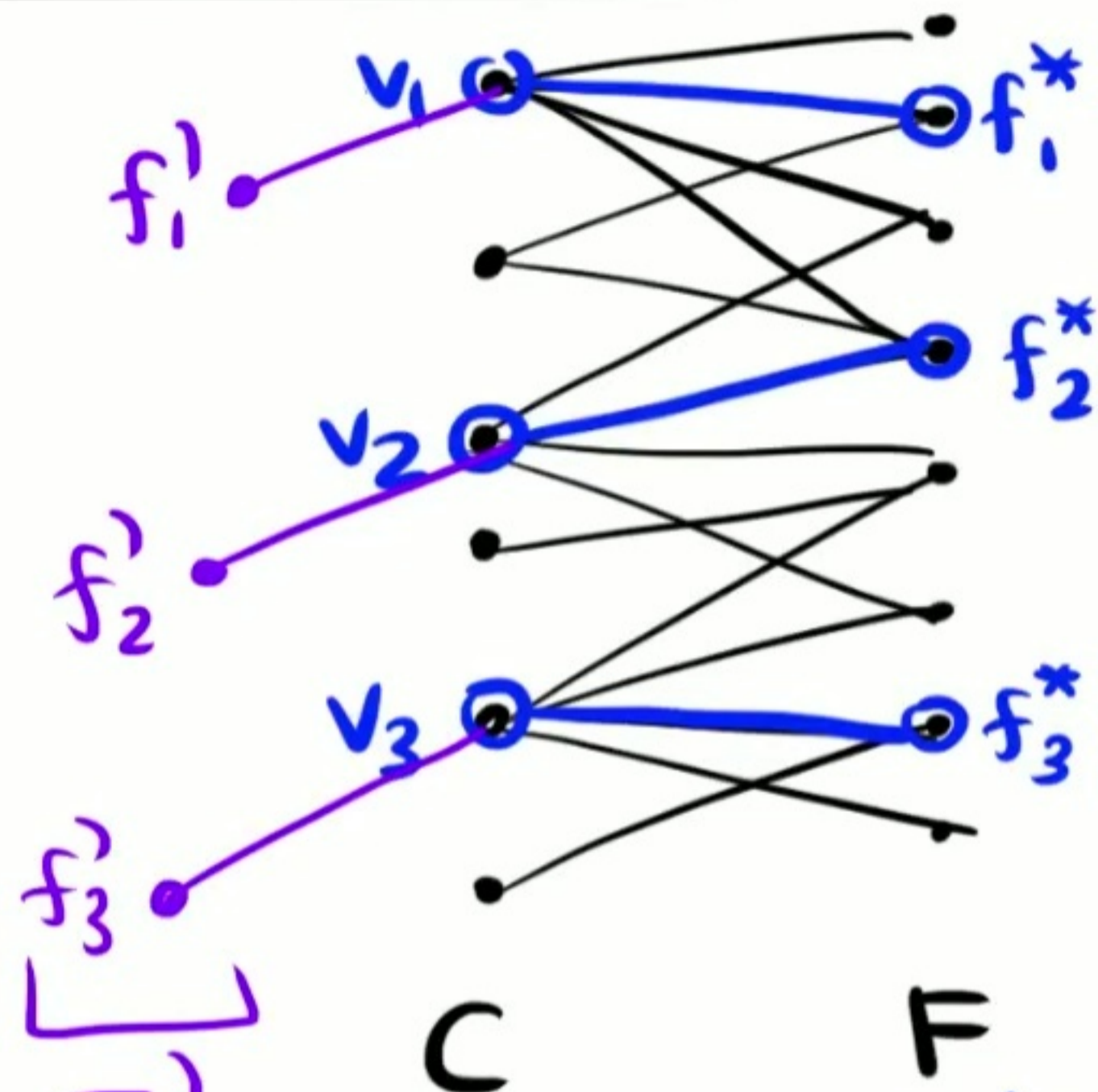
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"diminishing marginal returns"

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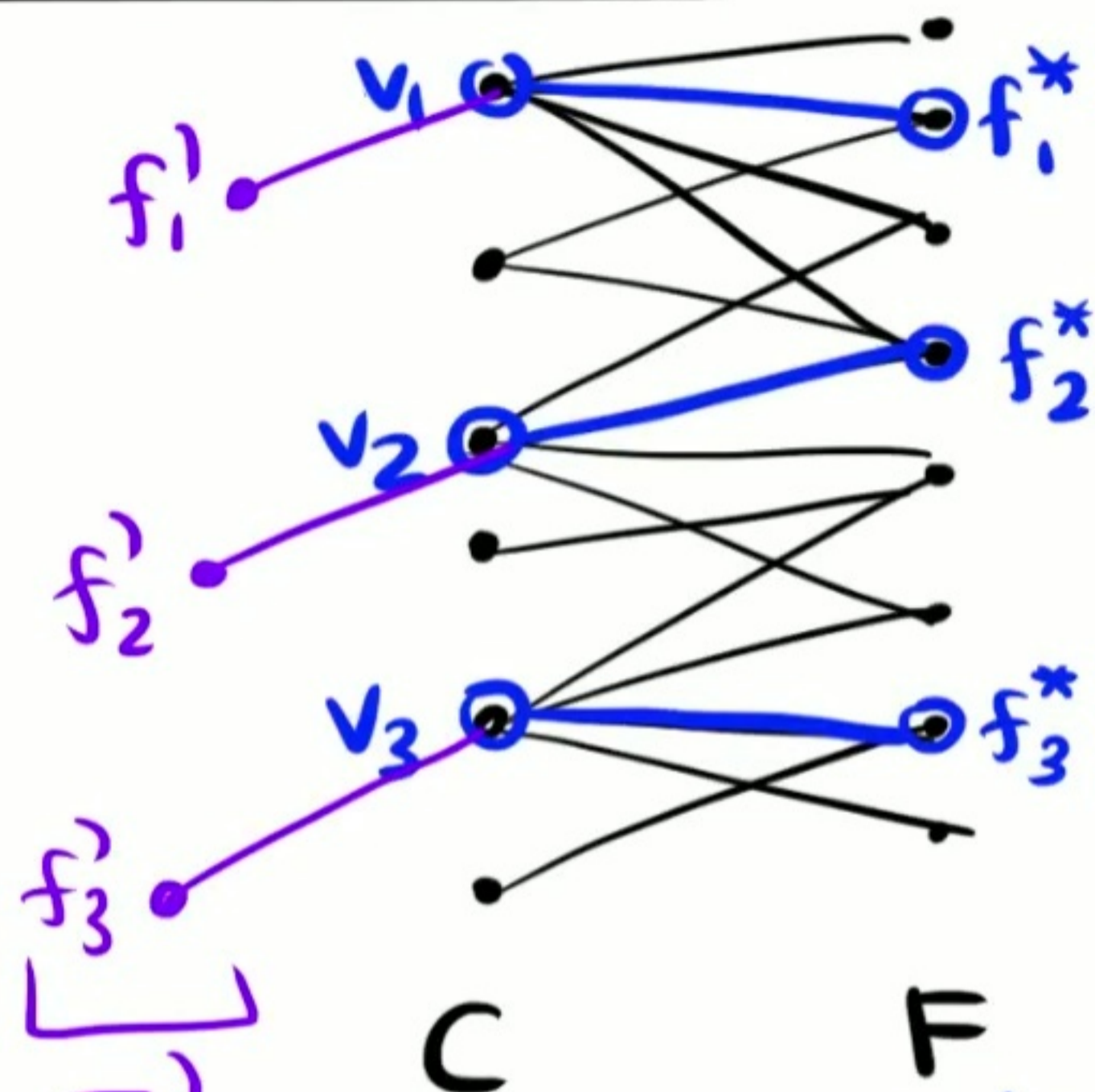
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 $\Leftrightarrow \forall v: -d(v, F' \cup S \cup f) + d(v, F' \cup S) \geq -d(v, F' \cup T \cup f) + d(v, F' \cup T)$



"diminishing marginal returns"

Optimization function

maximize

$$\text{impr}(S) := \sum_v w(v) (d(v, F') - d(v, F' \cup S))$$

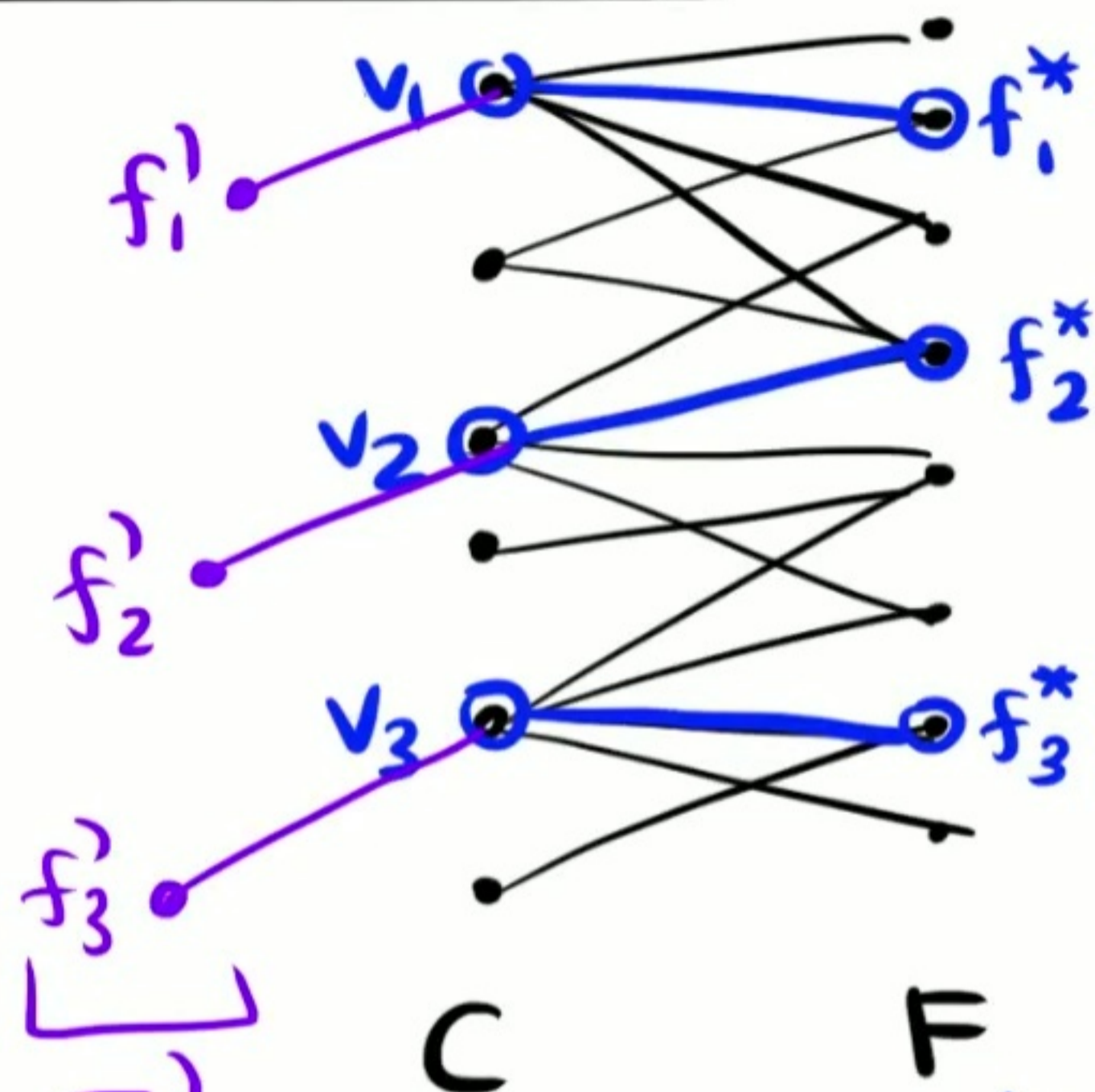
over $S \subseteq F, S = \{f_1, \dots, f_k, f_i \in N(v_i)\}$

Lemma: $\text{impr}(S)$ is ① monotone ③ submodular with ② $\text{impr}(\emptyset) = 0$.

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Optimization function

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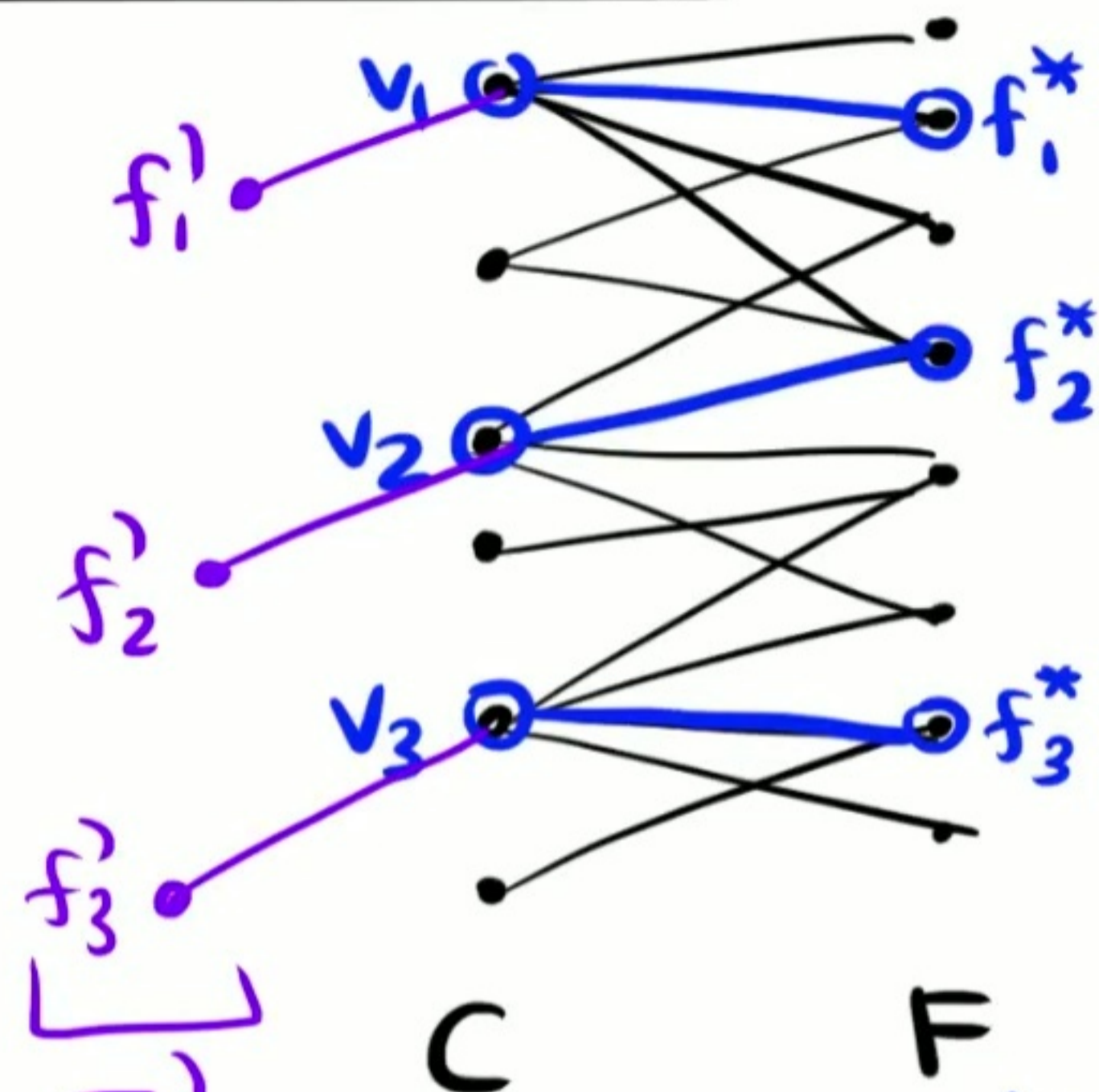
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"diminishing marginal returns"

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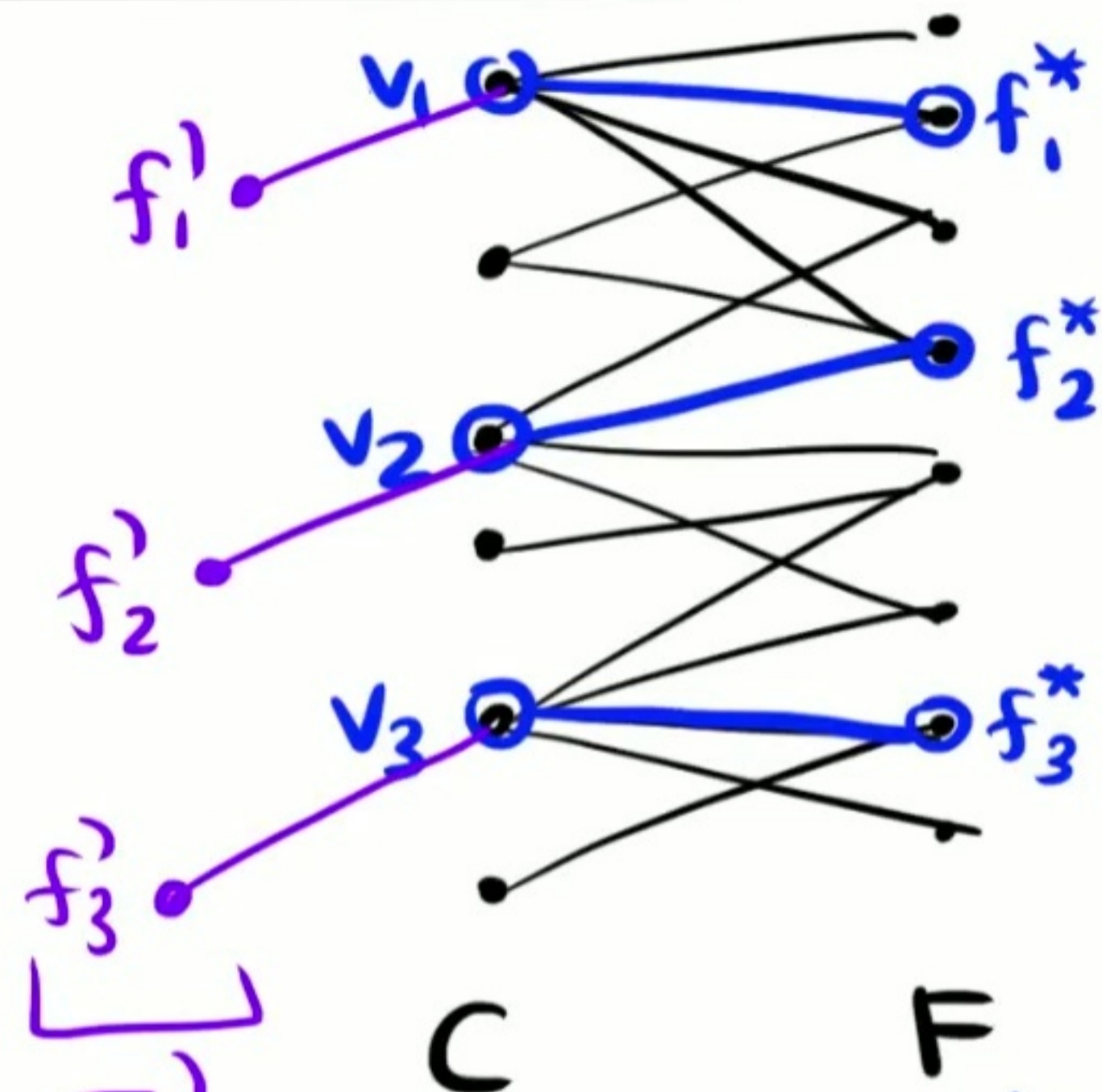
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"diminishing marginal returns"

Optimization function

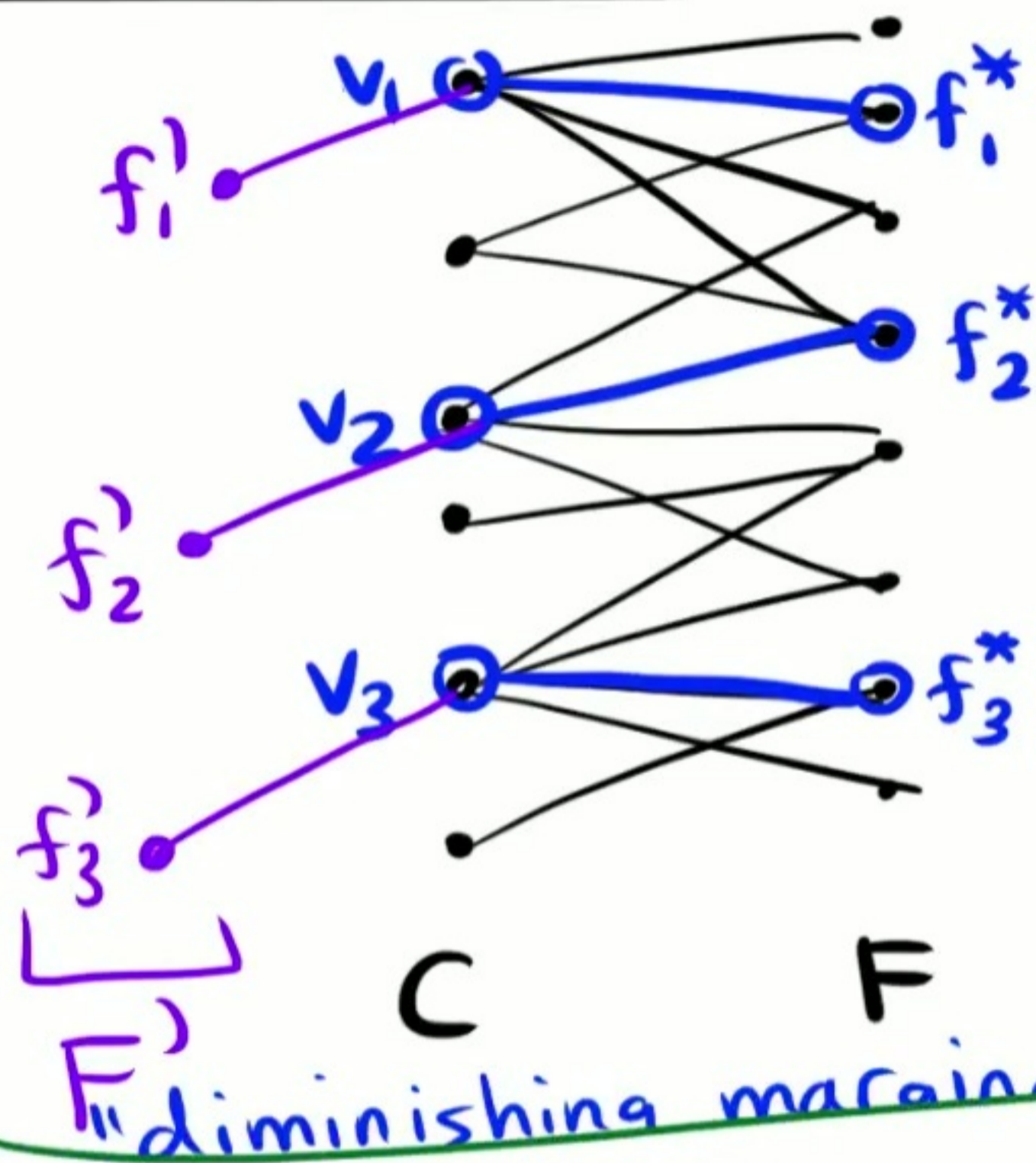
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Thm [CCPV] can $(1 - 1/e)$ -approximate
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Optimization function

maximize

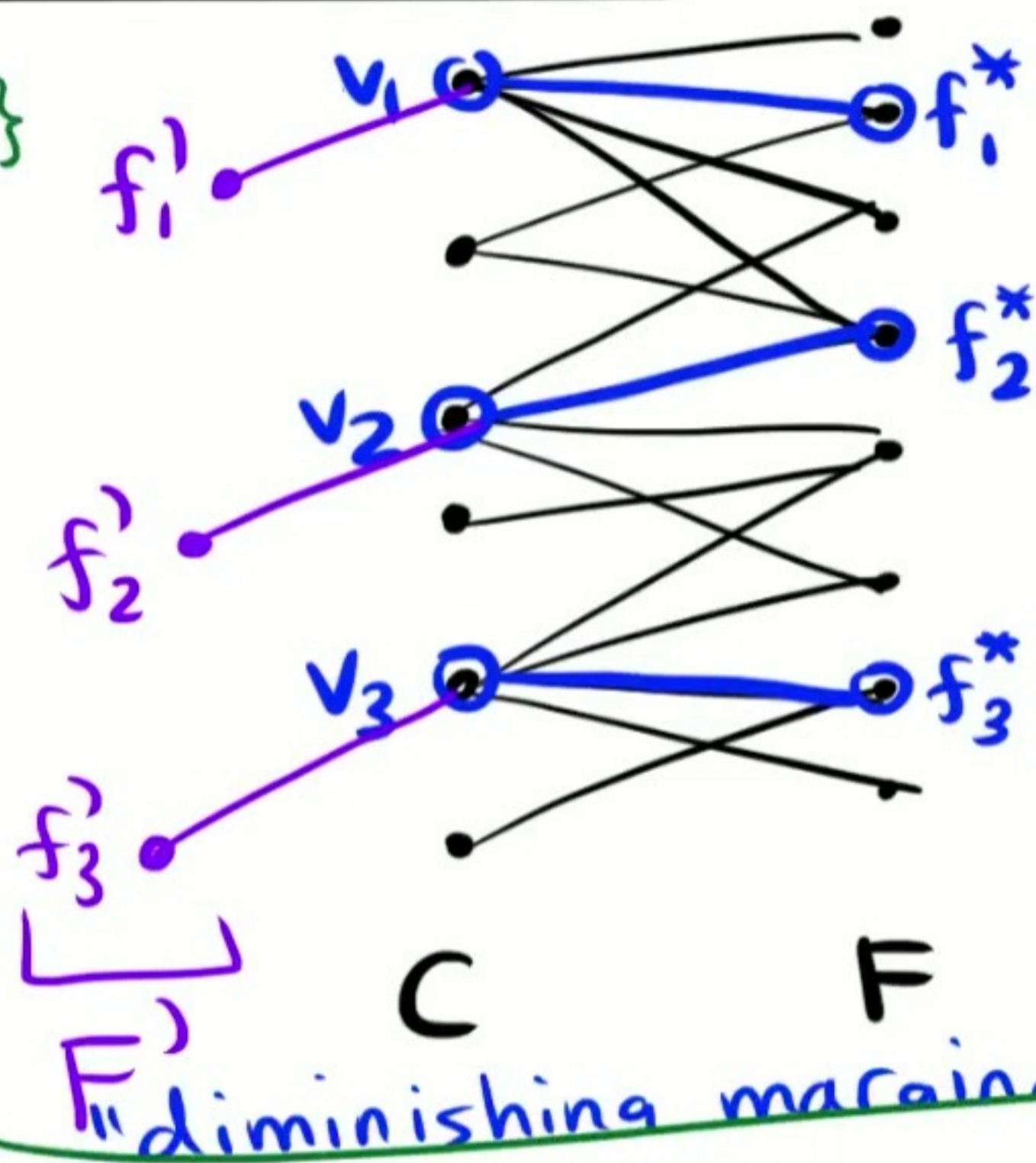
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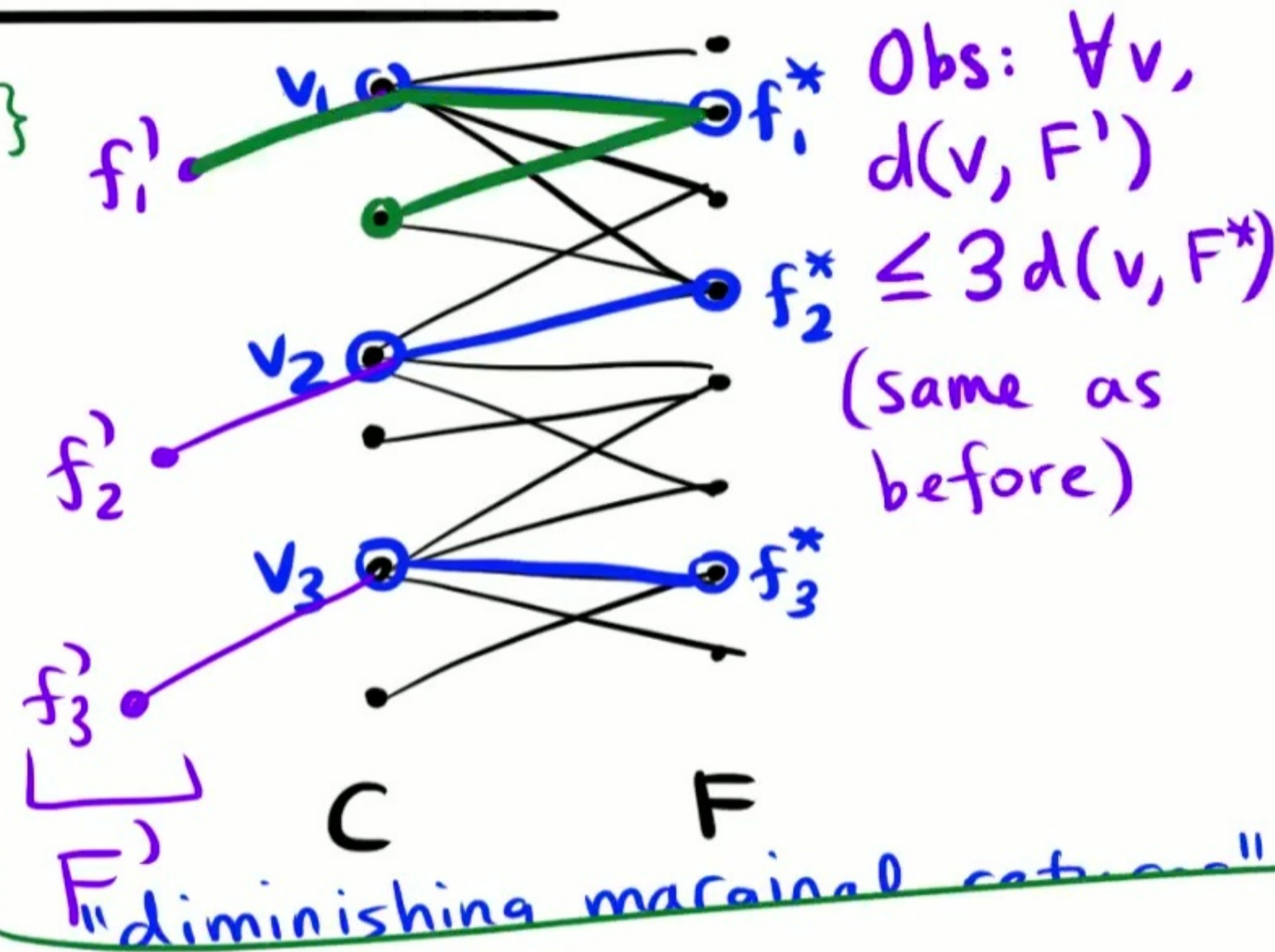
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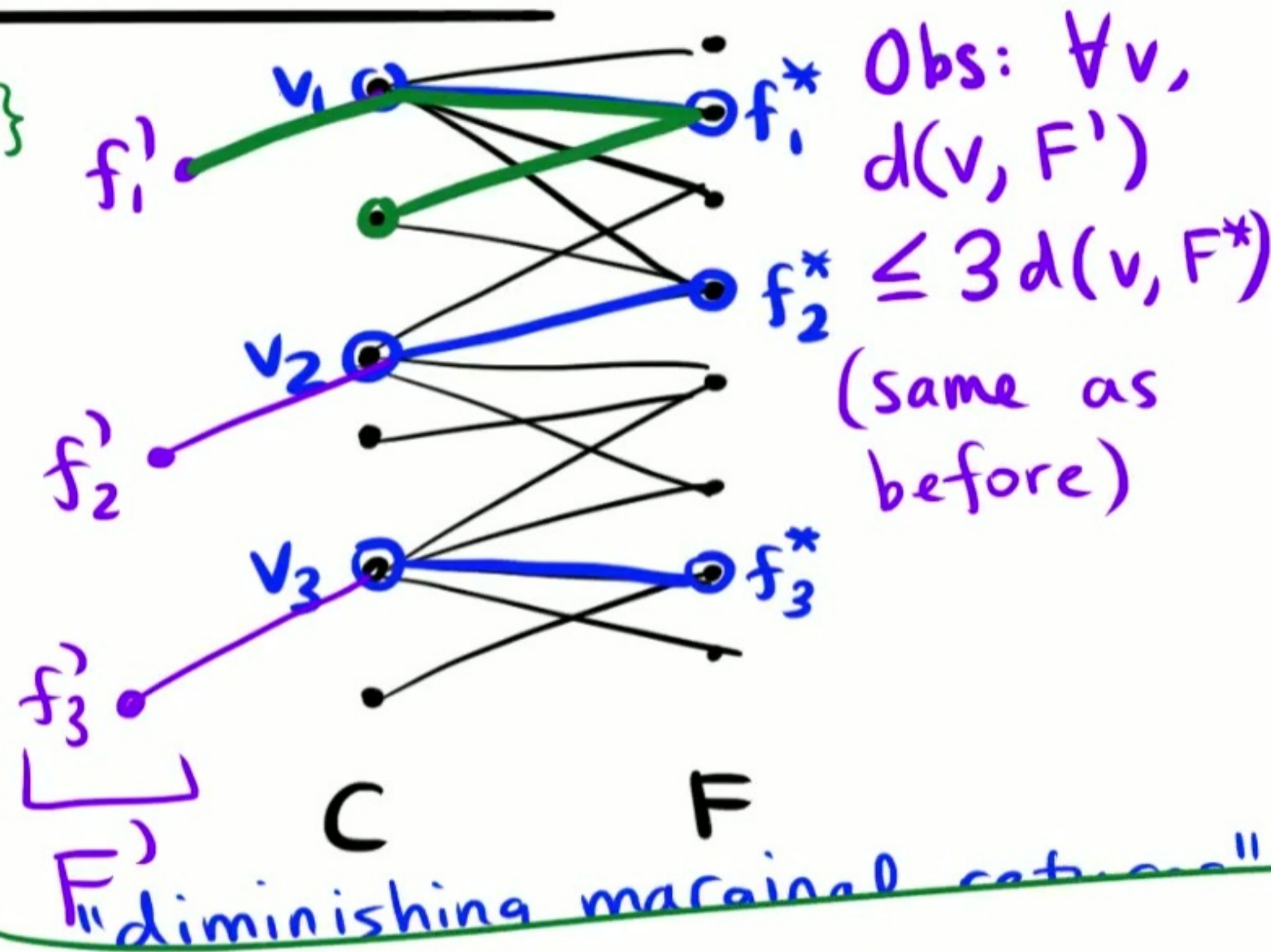
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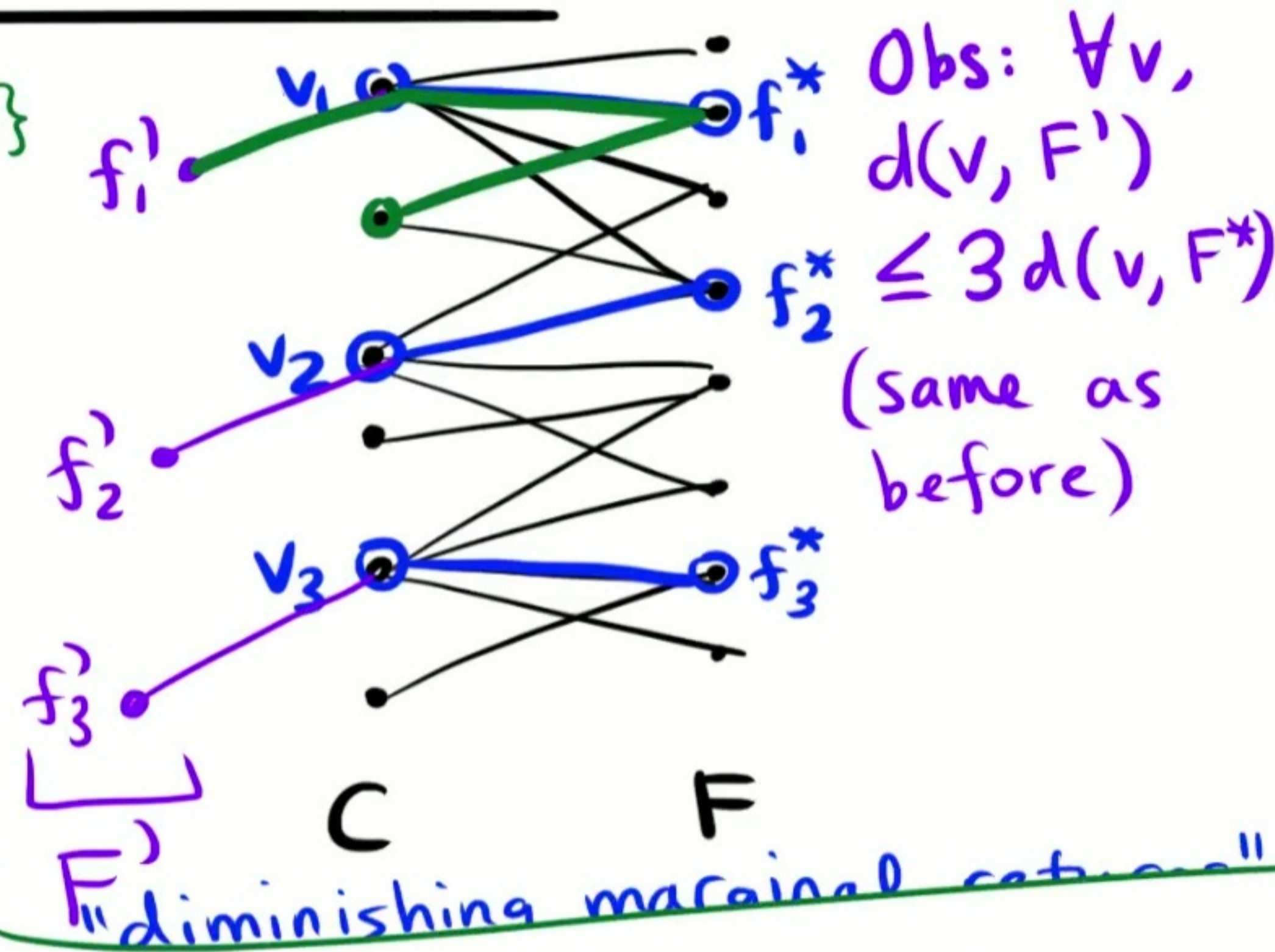
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Optimization function

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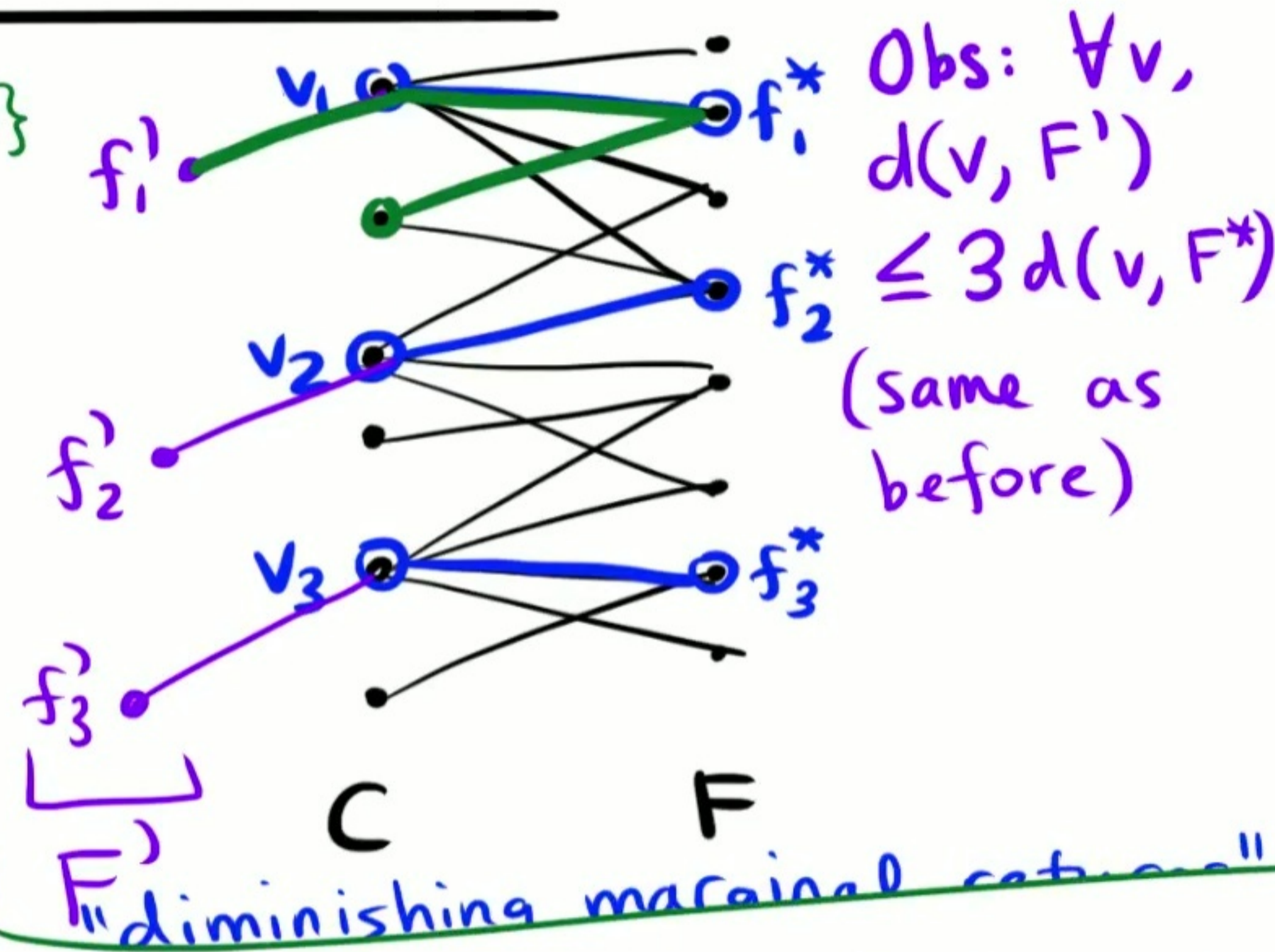
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Optimization function

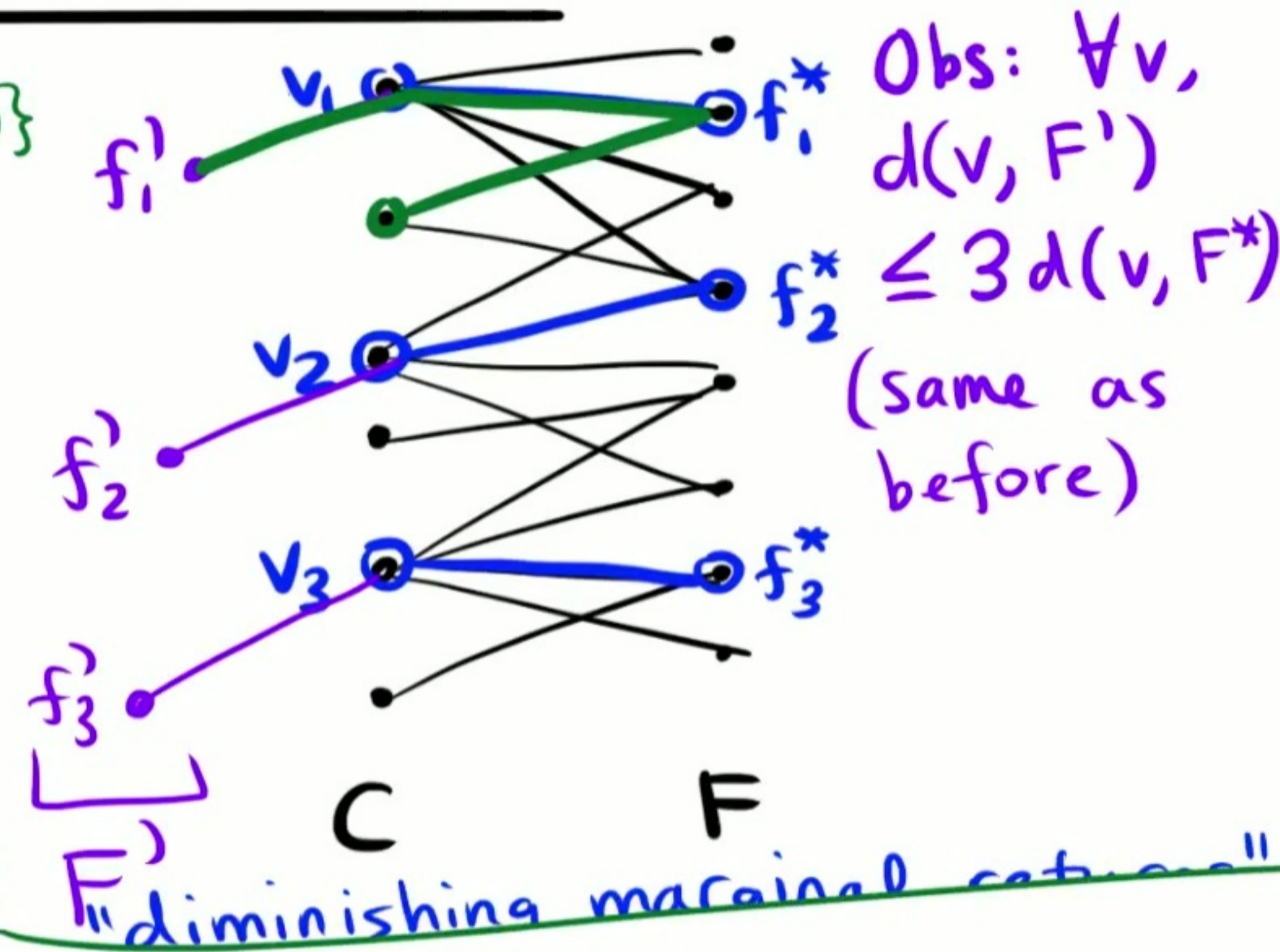
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Optimization function

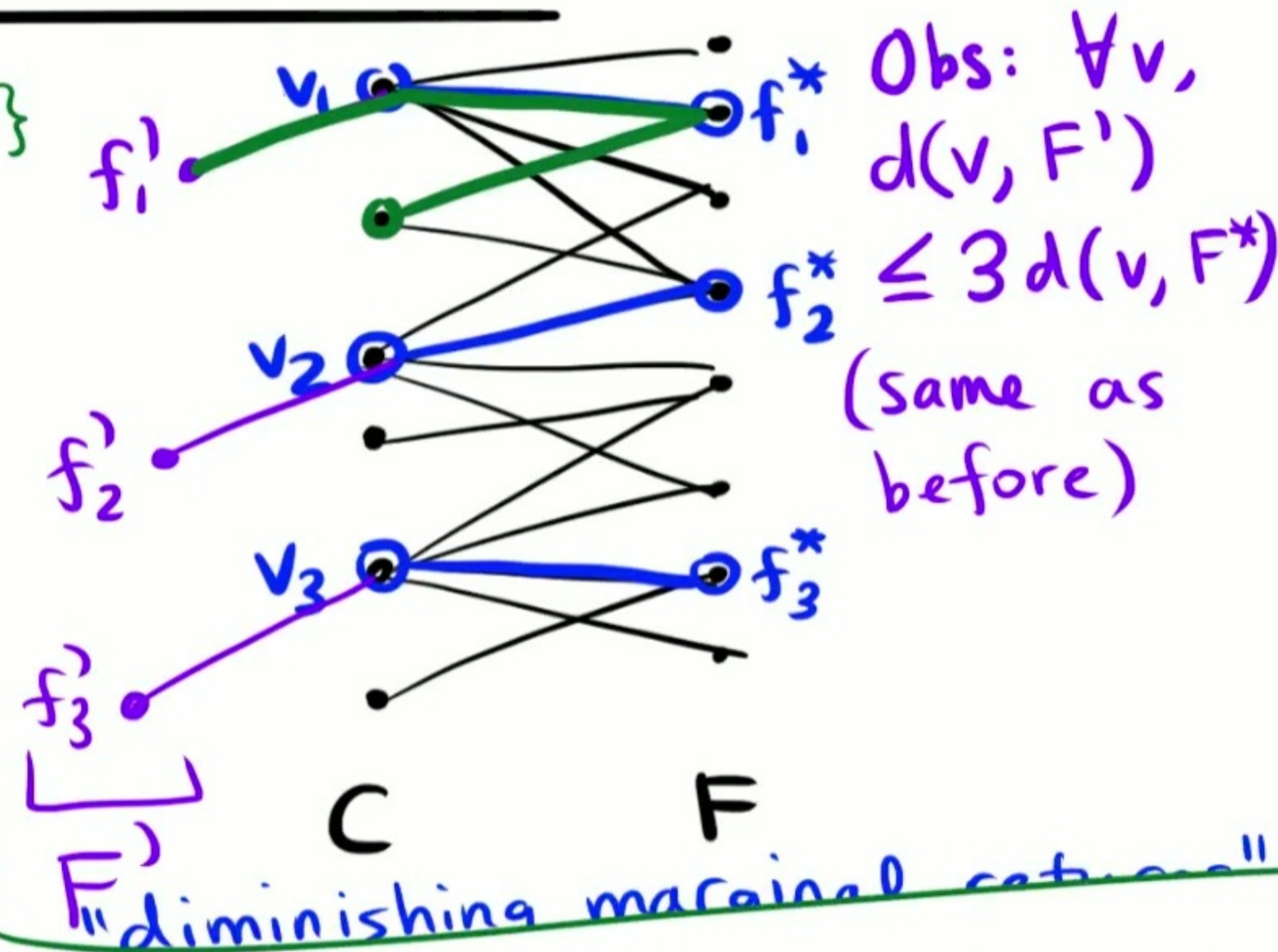
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Optimization function

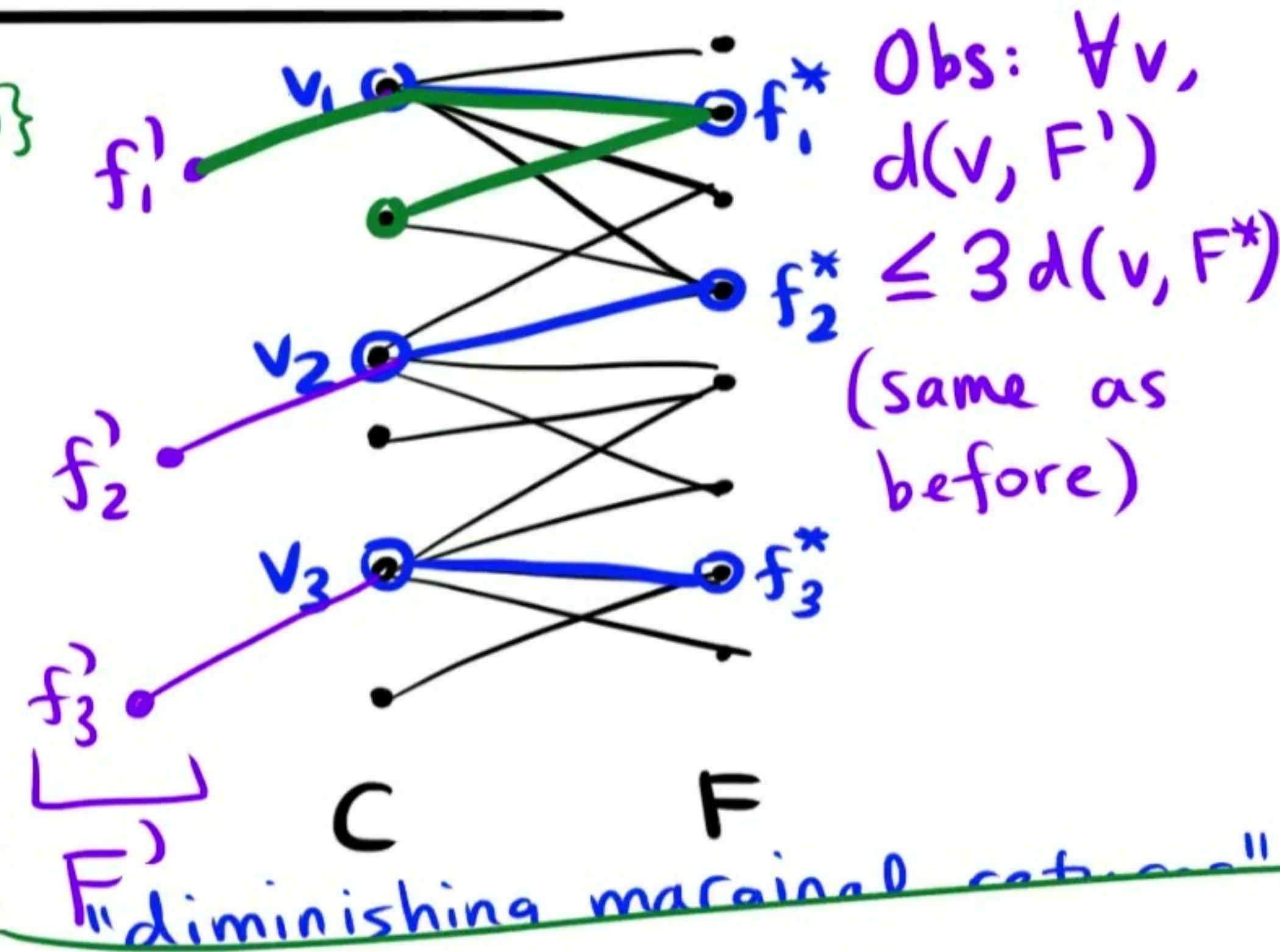
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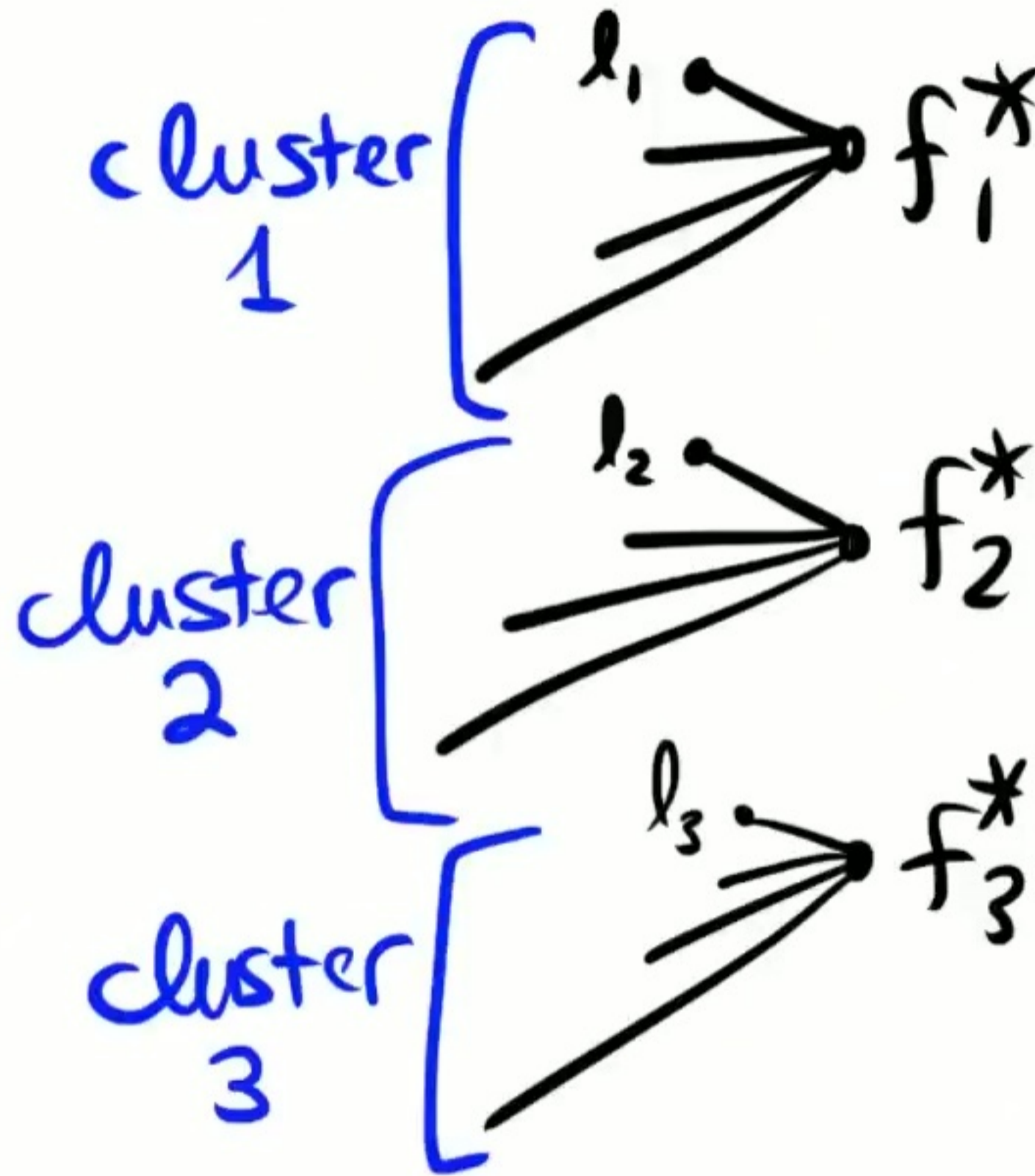
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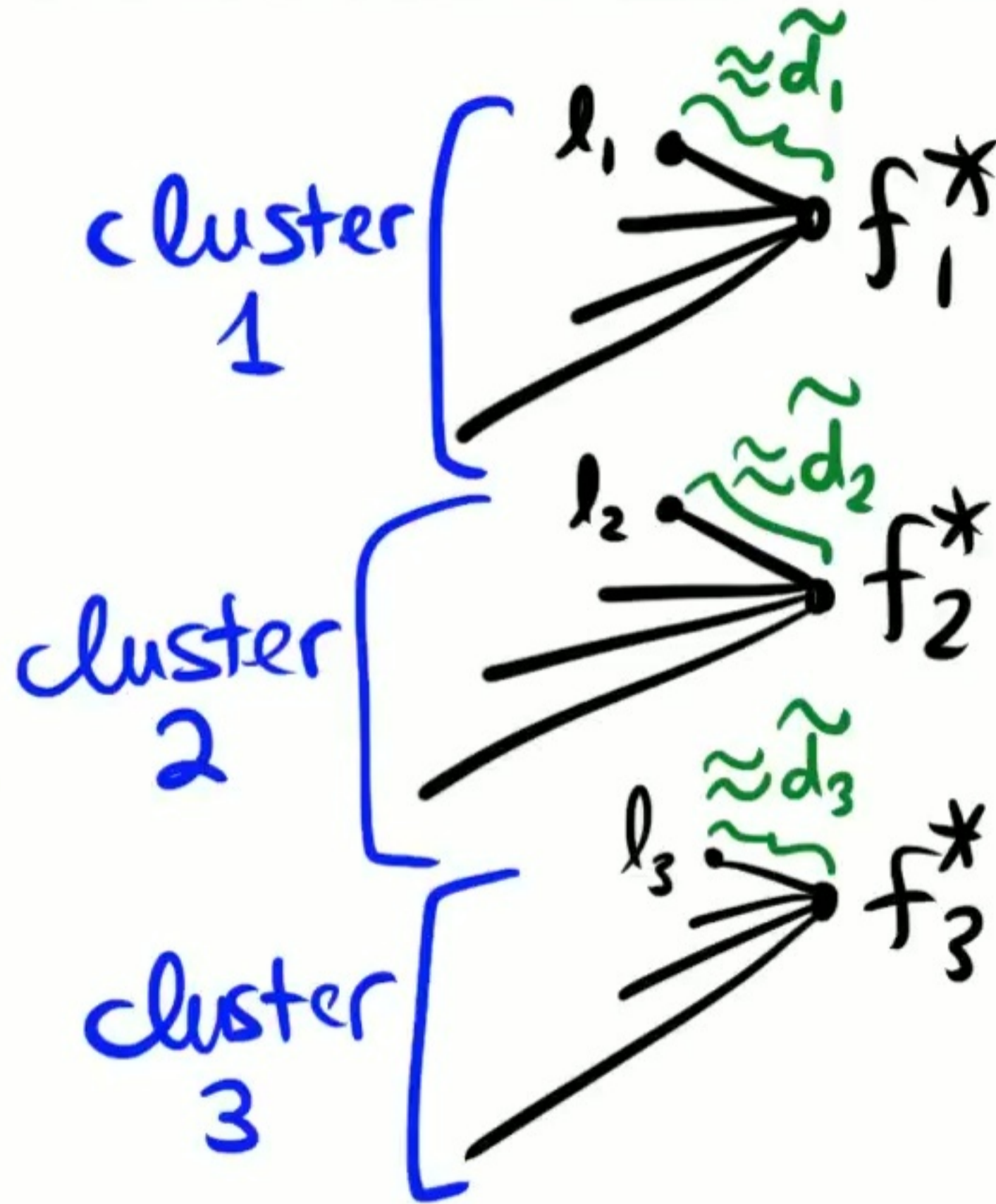
General case

- cluster i : all clients assigned to f_i^* in OPT
- l_i := closest client to f_i^* in cluster i



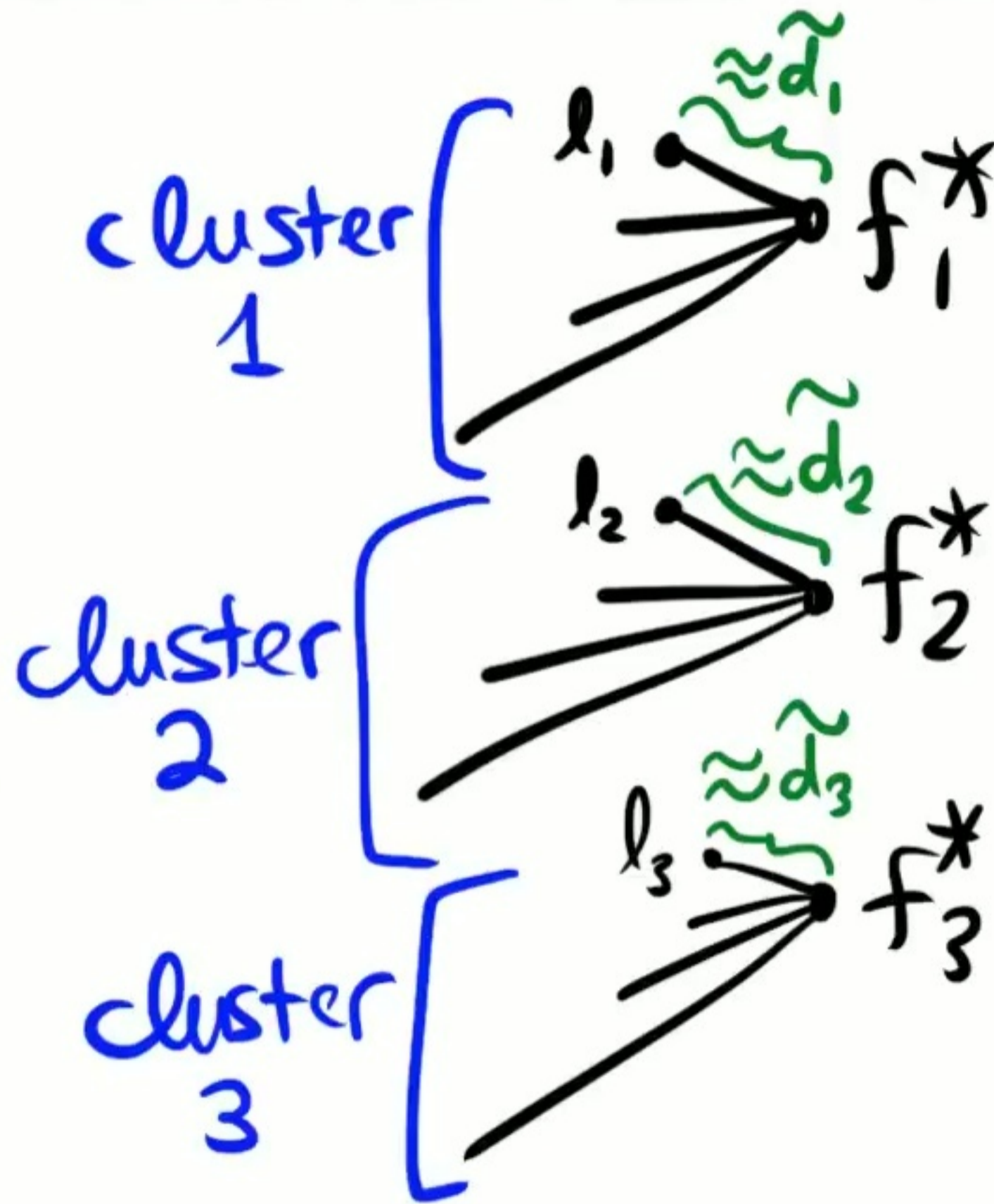
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- Guess $\tilde{d}_i \approx d(l_i, f_i^*)$ up to factor $(1+\epsilon)$



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- Define $N(l_i) := \{f \in F : d(l_i, f) \underset{(1+\epsilon)}{\approx} \tilde{d}_i\}$
[so $f_i^* \in N(l_i)$]



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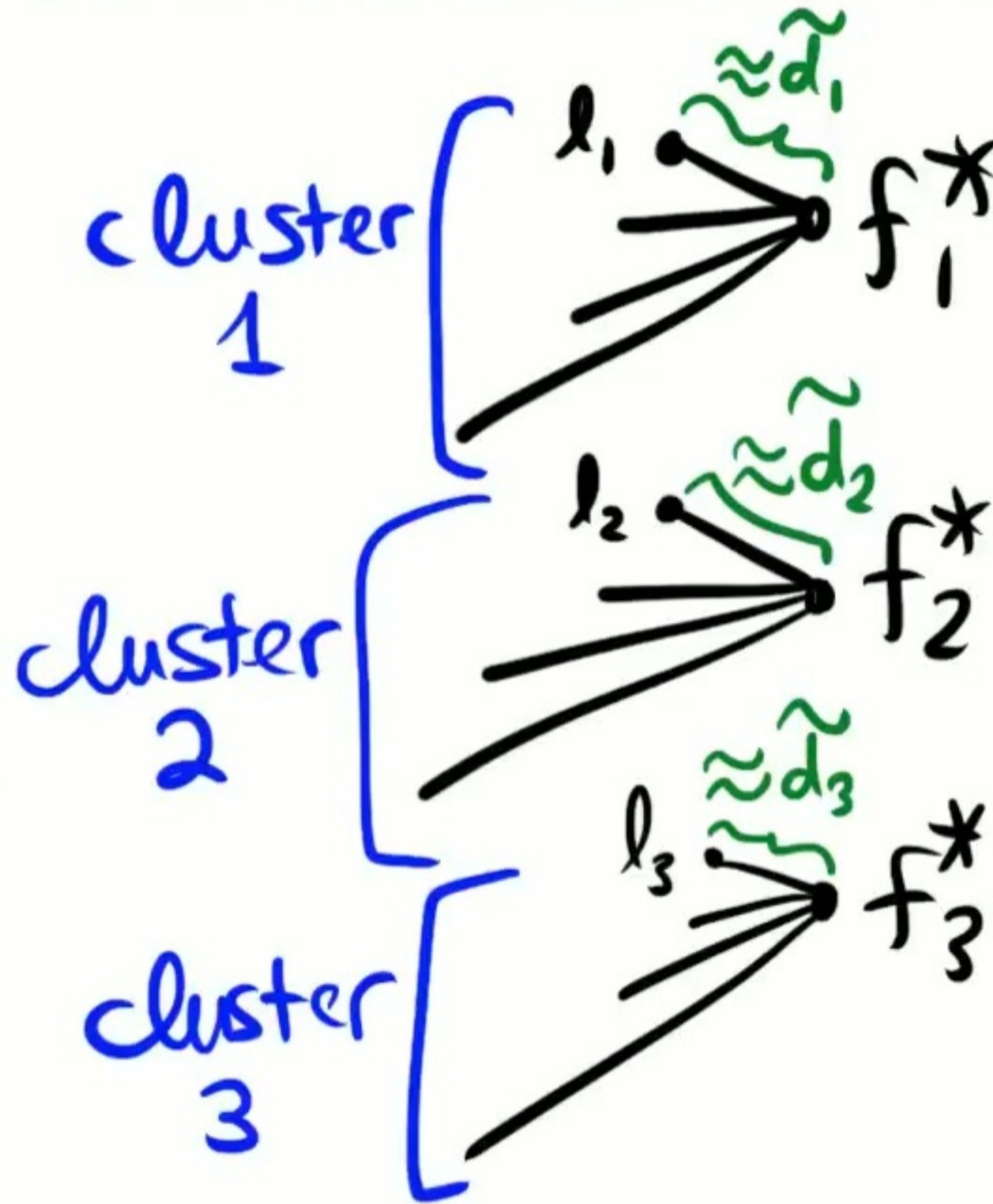
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• $d(v, F') \leq 3d(v, F^*)$

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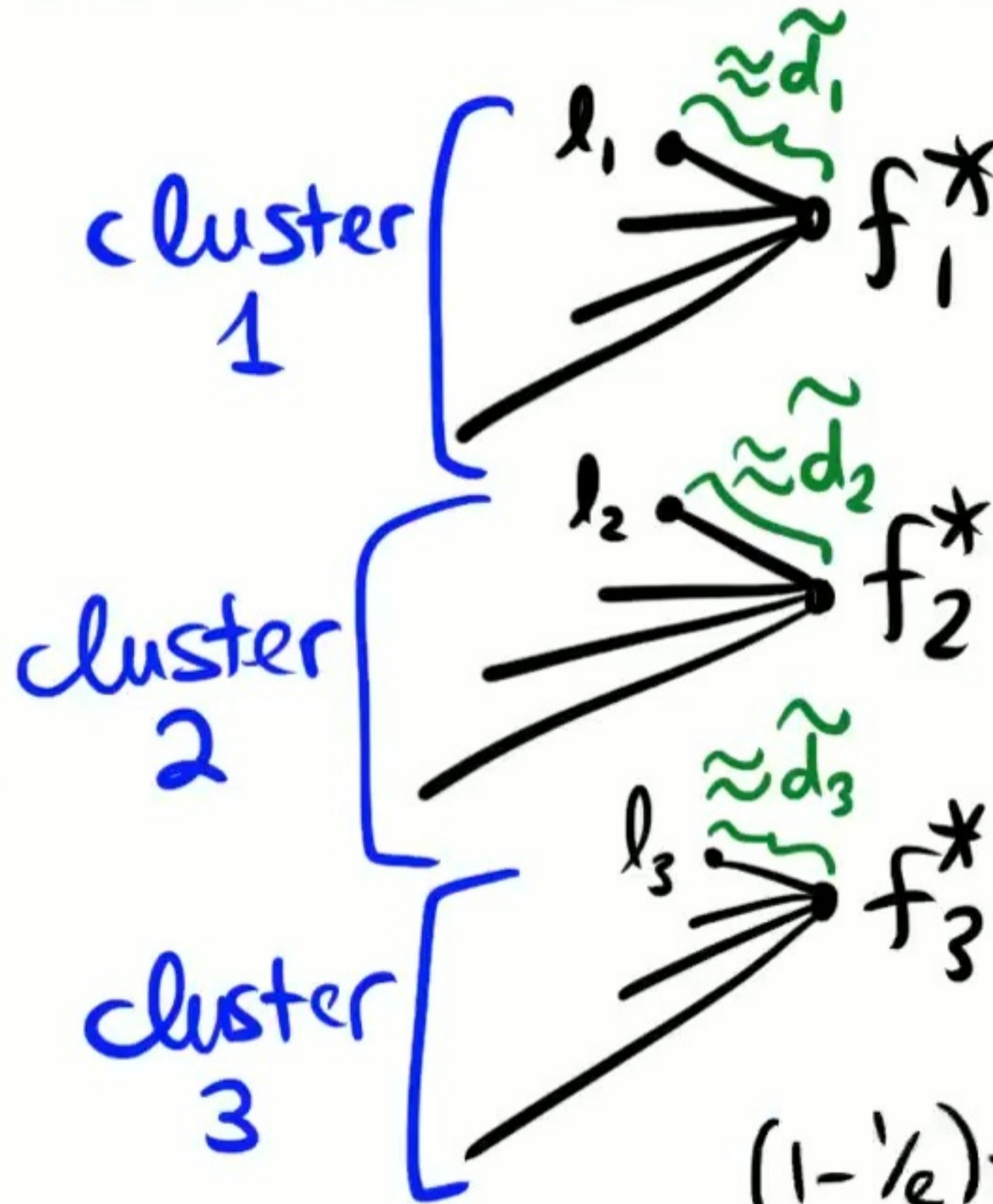
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$(1 - 1/e)$ -apx to $\text{impr}(S)$

$\Rightarrow (1 + 2/e + \epsilon)$ -apx to k -median.

Open problems

- Can FPT techniques be used for polytime k -median?
 - Fallback of 3-approx, submod opt
- Other variants?
 - Capacitated k -median: upper bound $3+\epsilon$
Lower bound $1+2/e$
 - Matroid median: upper bound $2+\epsilon$
Lower bound $1+2/e$