

# League of Legends Ban–Pick as a Min-Cost Flow Problem

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Draft pick, but make it linear programming.

# Motivation: Ban–Pick as an Optimization Problem

## Why model ban–pick?

- Ban–pick phase determines the lineup before the game starts.
- Conceptual data we have:
  - Roles  $R = \{\text{TOP}, \text{JGL}, \text{MID}, \text{ADC}, \text{SUP}\}$ ,
  - Available champions  $C$  and bans  $B$ ,
  - Utilities  $u_{r,c}$  for champion  $c$  on role  $r$ .
- Question: given  $u_{r,c}$ , bans, and eligibility, what is the optimal lineup for our team?
- We encode this as a structured min-cost flow problem.



LoL champion select screen (one lineup per role, no duplicates).

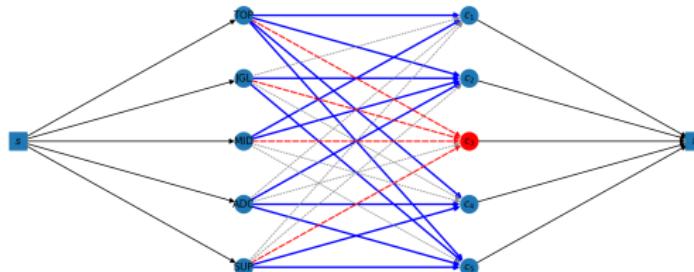
# Model Ingredients & Network Structure

## Data

- Roles  $R = \{\text{TOP}, \text{JGL}, \text{MID}, \text{ADC}, \text{SUP}\}$ ,  $|R| = 5$ .
- Champion set  $C \subseteq C_{\text{all}}$ ; team bans  $B \subseteq C$ .
- Eligibility  $1_{r,c} \in \{0, 1\}$ : 1 if champion  $c$  can be played on role  $r$ .
- Utility matrix  $u \in \mathbb{R}^{R \times C}$ ;  $u_{r,c}$  = performance of  $c$  on  $r$ .

## Flow network

- Nodes:  $V = \{s\} \cup R \cup C \cup \{t\}$ .
- Arcs:  $A_{sR} = \{(s, r) : r \in R\}$ ,  $A_{RC} = \{(r, c) : r \in R, c \in C, 1_{r,c} = 1, c \notin B\}$ ,  $A_{Ct} = \{(c, t) : c \in C\}$ ,  $A = A_{sR} \cup A_{RC} \cup A_{Ct}$ .
- Capacities:  $u_a = 1$  for all  $a \in A$ .
- Supplies/demands:  $b_s = |R|$ ,  $b_t = -|R|$ ,  $b_v = 0$  for  $v \notin \{s, t\}$ .
- Costs:  $c_{s,r} = 0$ ,  $c_{r,c} = -u_{r,c}$ ,  $c_{c,t} = 0$ .



Layered network: source–roles–champions–sink.

# Min-Cost Flow Formulation

## Decision variables

$$x_{s,r} \ (r \in R), \quad x_{r,c} \ ((r, c) \in A_{RC}), \quad x_{c,t} \ (c \in C), \quad 0 \leq x_a \leq 1 \ \forall a \in A.$$

## Objective

$$\min_x \sum_{(r,c) \in A_{RC}} (-u_{r,c}) x_{r,c} \iff \max_x \sum_{(r,c) \in A_{RC}} u_{r,c} x_{r,c}.$$

## Flow-balance constraints

$$\sum_{r \in R} x_{s,r} = |R|, \quad (\text{source})$$

$$\sum_{c: (r,c) \in A_{RC}} x_{r,c} - x_{s,r} = 0, \quad \forall r \in R \ (\text{roles})$$

$$x_{c,t} - \sum_{r: (r,c) \in A_{RC}} x_{r,c} = 0, \quad \forall c \in C \ (\text{champions})$$

$$\sum_{c \in C} x_{c,t} = |R|, \quad (\text{sink}).$$

## Interpretation

- Exactly  $|R|$  units of flow go from  $s$  to  $t$ .
- Unit capacities enforce one champion per role and no duplicate champions.

# Toy Example: Optimal Lineup

## Toy utility matrix

	G	D	A
TOP	8	7	4
MID	5	3	9

## Feasible lineups (no duplicates)

- (TOP G, MID A): total utility  $8 + 9 = 17$ .
- (TOP D, MID A): total utility  $7 + 9 = 16$ .
- (TOP G, MID D): total utility  $8 + 3 = 11$ .

## Min-cost flow solution

- Chooses (TOP G, MID A).
- Optimal total utility = 17, min total cost = -17.
- Corresponds to 2 units of integral flow from  $s$  to  $t$  via G and A.

# Parametric Buff: When Does TOP Switch Champions?

## Parametric setup

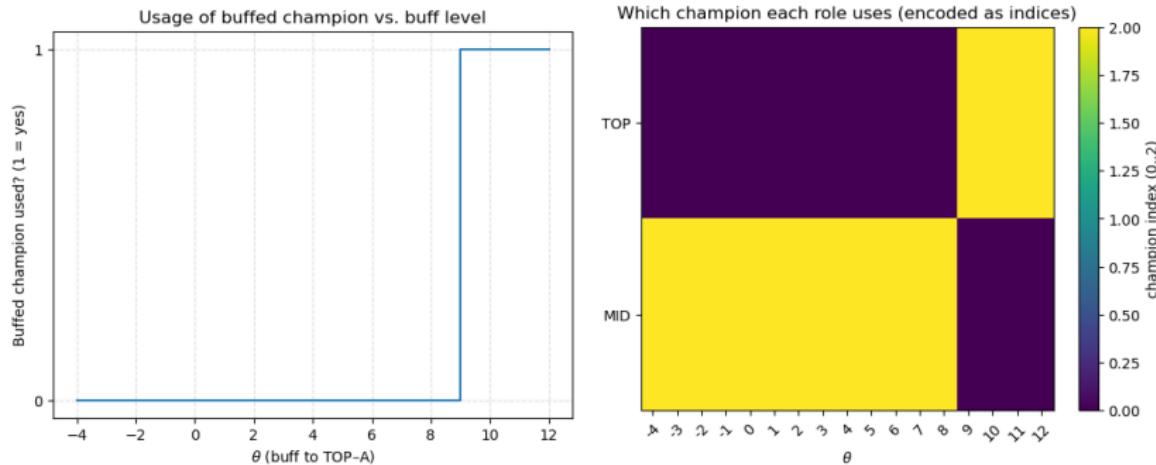
- Fix all utilities except TOP-A:

$$u_{\text{TOP},A}(\theta) = u_{\text{TOP},A}^{(0)} + \theta.$$

- For each  $\theta$ , update costs  $c_{r,c}(\theta) = -u_{r,c}(\theta)$  and re-solve the same min-cost flow model.

## Behavior in the toy example

- For  $\theta \leq 8$ : optimal lineup is (TOP G, MID A), total utility = 17.
- For  $\theta > 8$ : lineup switches to (TOP A, MID G), total utility =  $9 + \theta$ .
- The value function  $-Z^*(\theta)$  is piecewise linear; the lineup changes exactly at  $\theta = 8$ .



Usage of the buffed champion (left) and role-swap heatmap (right).

# Summary & Outlook

## Summary

- Modeled LoL ban–pick for a single team as a structured min-cost flow problem.
- Layered network  $s \rightarrow R \rightarrow C \rightarrow t$  with unit capacities enforces “one champ per role, no duplicates”.
- Objective: choose role–champion assignments that maximize total utility.
- Parametric buff analysis yields piecewise-linear value functions and clear thresholds where the optimal lineup changes.

## Implementation & extensions

- Implementable with standard LP / min-cost flow solvers in Python.
- Natural extensions: two-team game, sequential bans/picks, and game-theoretic models (extensive-form, Stackelberg, etc.).

Thanks! Questions or champion balance ideas?

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