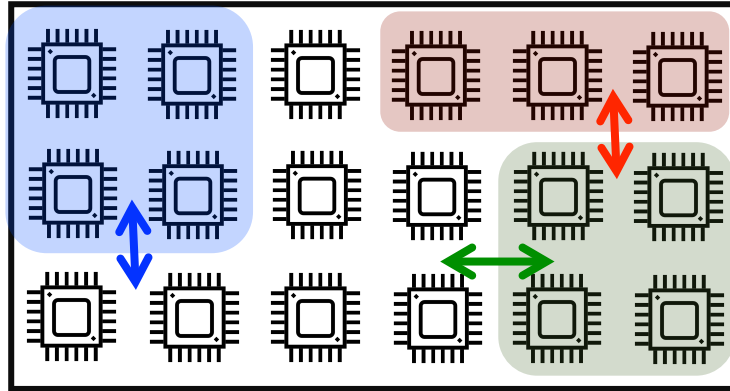


# Algorithms for Cloud Computing

## part II: Resource Allocation and Fairness

**Georgios Paschos**

# Resource Allocation



Sharing resources among users

- Each user receives a “satisfaction” from resources
  - Maximize total satisfaction with available resources
  - Caveat: might be *unfair*...

# Outline

- **Welfare maximization:** find resource allocations for the system's benefit
- **Fairness** and its relation to welfare maximization
- **Multi-resource fairness**

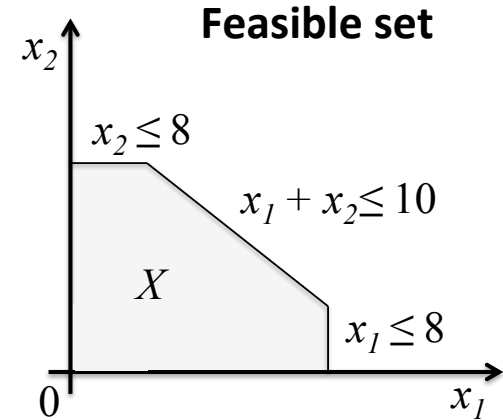
# Welfare Maximization

- **Problem formulation:**

- $K$ : number of users
- $x_k$ : allocated resources to user  $k$
- $X$ : set of feasible allocations (convex, bounded in positive orthant)
- $U_k$ : utility of user  $k$  (increasing, concave)

$$\max_{x \in \mathcal{X}} \sum_{k=1}^K U_k(x_k)$$

- **Convex solvers:** project gradient, Lagrangian relaxation, dual ascent, ADMM, ...



# Why care about welfare maximization?

- **Question:** why not just sell resources to the highest bidder?

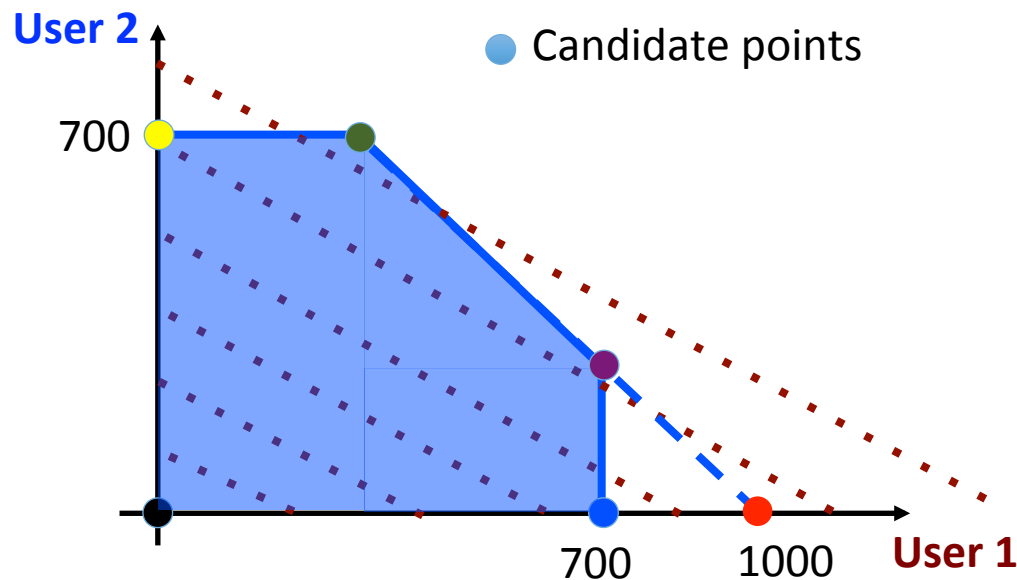
A1: In some systems, we sell SLAs and use available resources to meet them.  
Welfare helps to distribute surplus, and/or de-risk failing SLAs

A2: Sometimes SLAs can not be met, how do we decide which ones to violate, and by how much?

A3: Even if you do sell, to sell everything and not loose money, you still need to understand welfare maximization

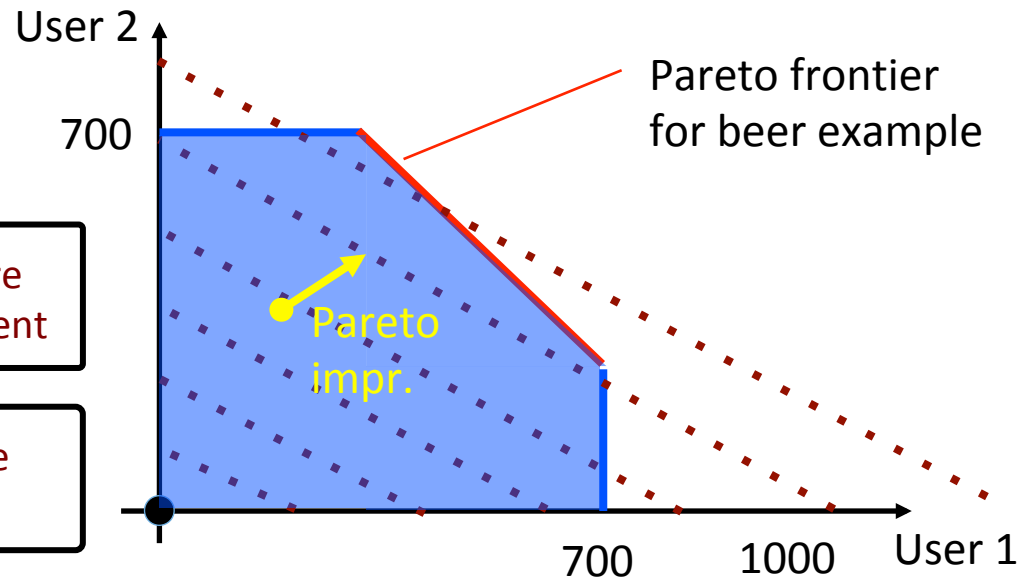
# The beer example

- We have 1lt of beer, two glasses of 700ml, and user 2 is twice “thirsty” than user 1.  $U(x) = x_1 + 2x_2$
- Questions: Describe feasible set. What point maximizes welfare?



# Pareto Efficiency

- **Definition:** A feasible allocation  $y$  is a Pareto improvement for  $x$  if  $U_k(y_k) \geq U_k(x_k)$ ,  $k = 1, \dots, K$ , and  $>$  for at least one
- **Definition:** A point is Pareto efficient if there is no Pareto improvement for it



**Lem:** The solution of welfare maximization is Pareto efficient

**Q:** what other points in the example can be Pareto?

# Questions

- Two persons, 100 bananas. Characterize the Pareto frontier
- Two persons, 2 bananas, 2 apples. P1 likes bananas and dislikes apples, P2 the opposite. Frontier?
- Two persons, 2 bananas, 2 apples. P1 likes bananas and is indifferent to apples, P2 the opposite. Frontier?
- P1 values 1 banana = 2 apples, and P2 the opposite. Frontier?

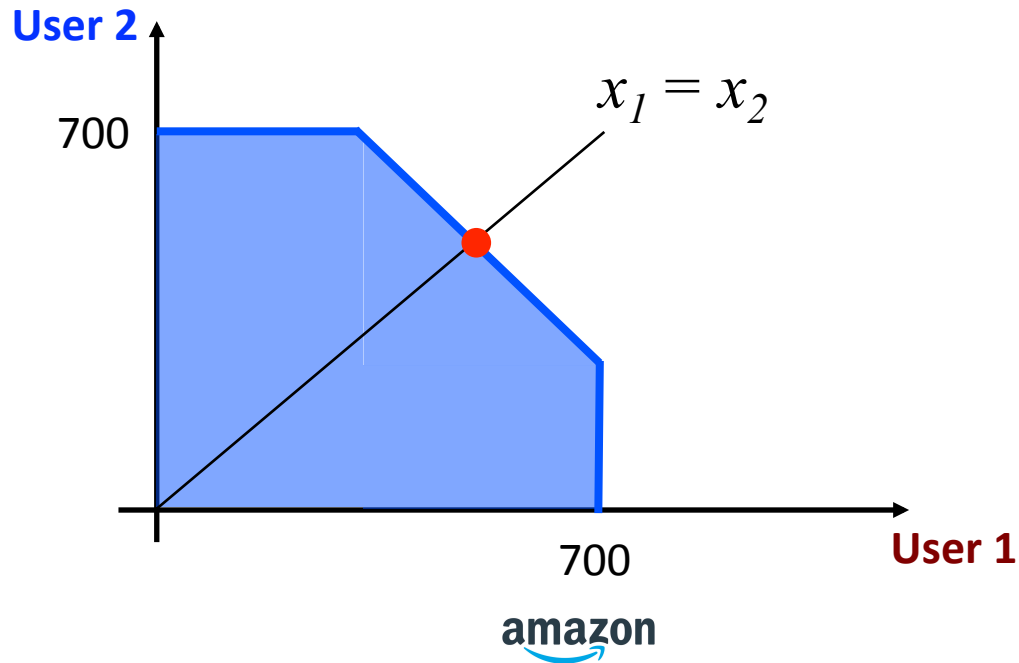


# Max-min fairness

- **Definition:** A feasible allocation  $x$  is Max-Min Fair (MMF) in set  $X$  if for any other  $y$  it holds:

$$y_m > x_m \Rightarrow \exists n \neq m : y_n < x_n \leq x_m$$

To improve the utility of user  $m$ , we must worsen the utility of a “poorer” user



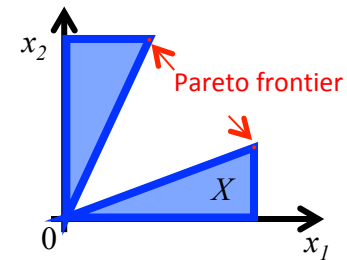
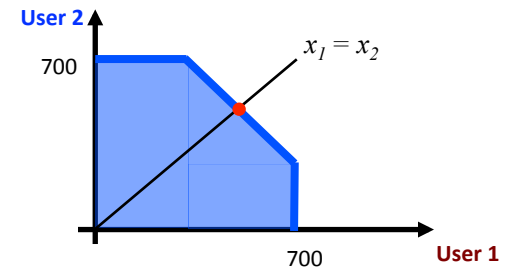
# Max-min fairness: claims

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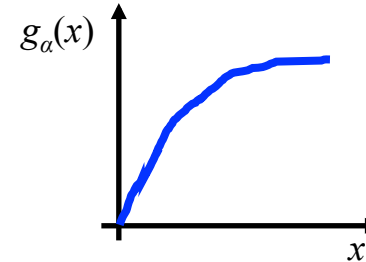
- If  $X$  is convex, there exists a unique MMF
- Non-convex sets might have no MMF (see example)
- MMF is equivalent to max the minimum and subject to that, max the 2<sup>nd</sup> min, etc..
- If “all equal” is Pareto, it is MMF



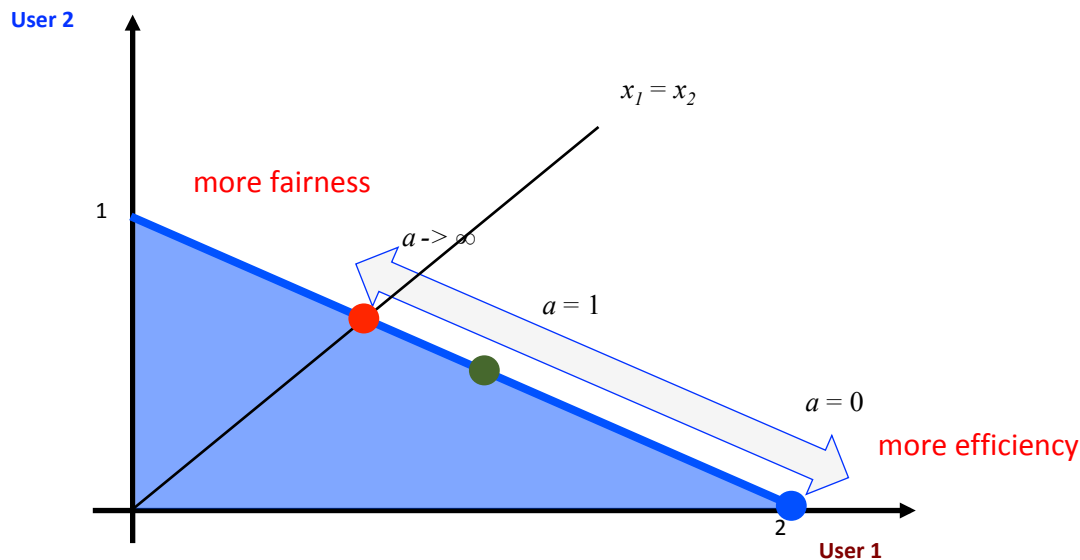
# Alpha Fairness

- Family of concave utility functions

$$g_\alpha(x) = \begin{cases} \frac{x^{1-\alpha}}{1-\alpha} & \text{if } \alpha \in [0, 1) \cup (1, \infty) \\ \log x & \text{if } \alpha = 1 \end{cases}$$

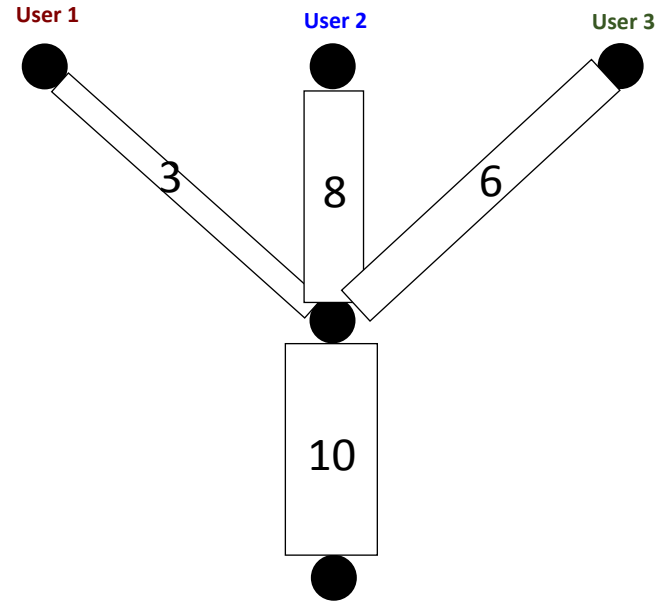


- Strictly convex Welfare  $\sum_{k=1}^K \frac{x_k^{1-\alpha}}{1-\alpha} \Rightarrow$  unique solution for  $\alpha > 0$



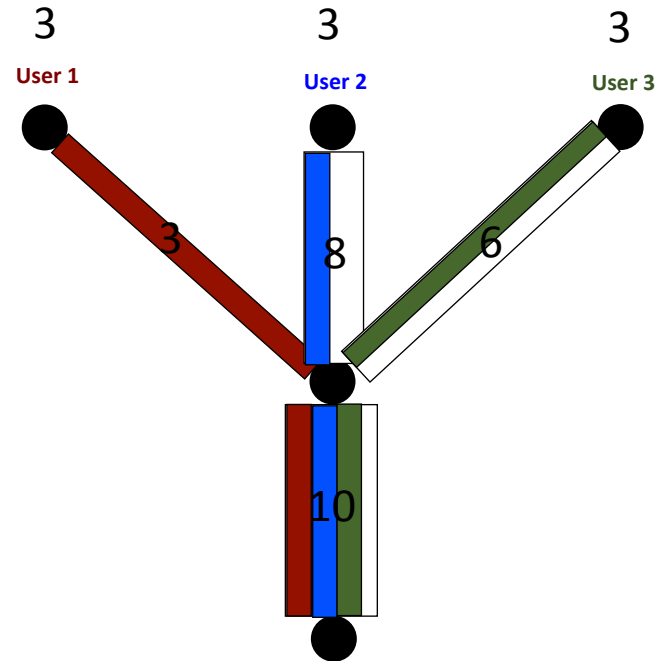
# Progressive Filling algorithm

- Increase iteratively allocation until reaching a bottleneck



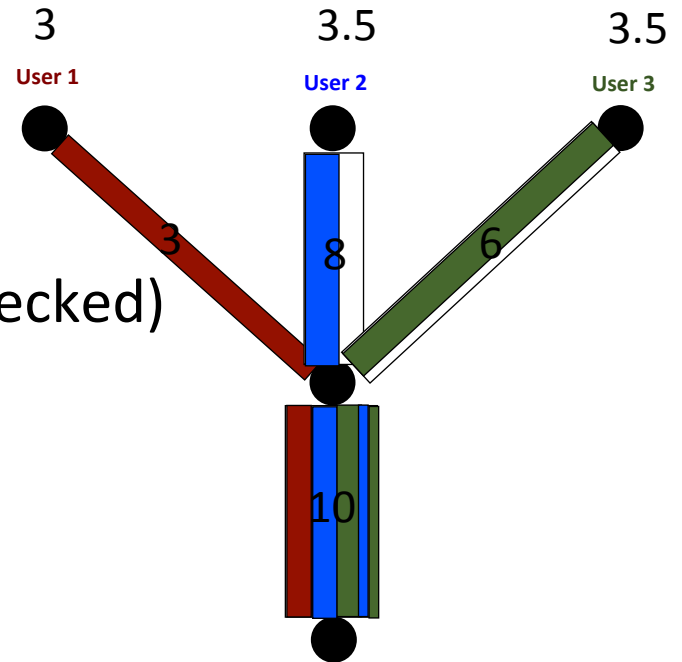
# Progressive Filling algorithm

- Increase iteratively allocation until reaching a bottleneck
- Step 1: fill up to 3
  - User 1 = 3 (bottlenecked)



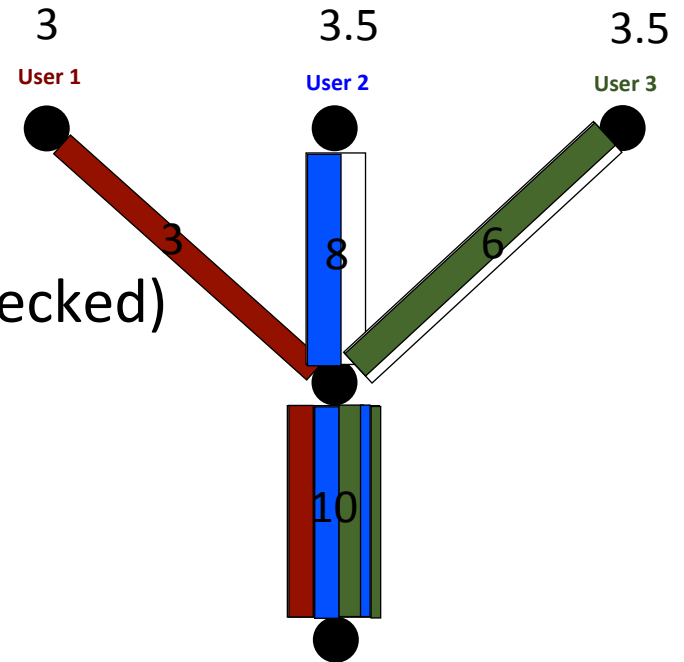
# Progressive Filling algorithm

- Increase iteratively allocation until reaching a bottleneck
- Step 1: fill up to 3
  - User 1 = 3 (bottlenecked)
- Step 2: fill up to 3.5
  - User 2 and User 3 = 3.5 (bottlenecked)
- Algorithm stops



# Progressive Filling algorithm

- Increase iteratively allocation until reaching a bottleneck
- Step 1: fill up to 3
  - User 1 = 3 (bottlenecked)
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  - User 2 and User 3 = 3.5 (bottlenecked)
- Algorithm stops

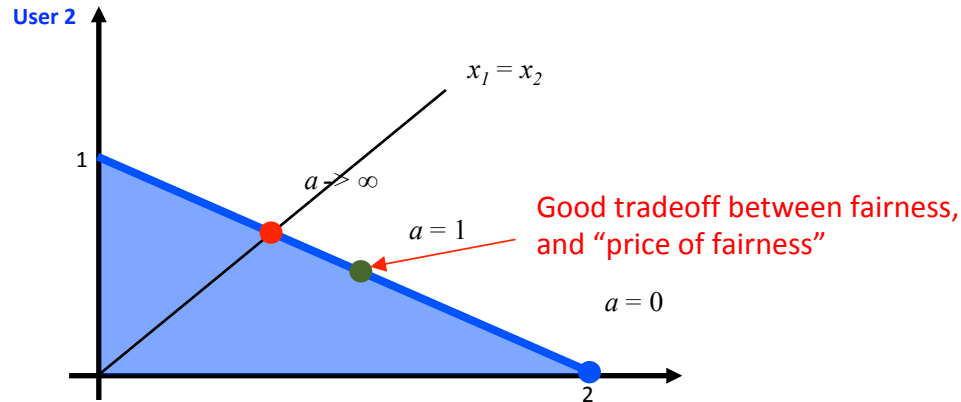


**Thm:** Progressive Filling converges to MMF.

# Proportional fairness ( $\alpha=1$ )

- **Definition:** A feasible allocation  $x$  is Proportional Fair (PF) in set  $X$  if any other  $y$  has a negative average change:  $\sum_k \frac{y_k - x_k}{x_k} \leq 0$

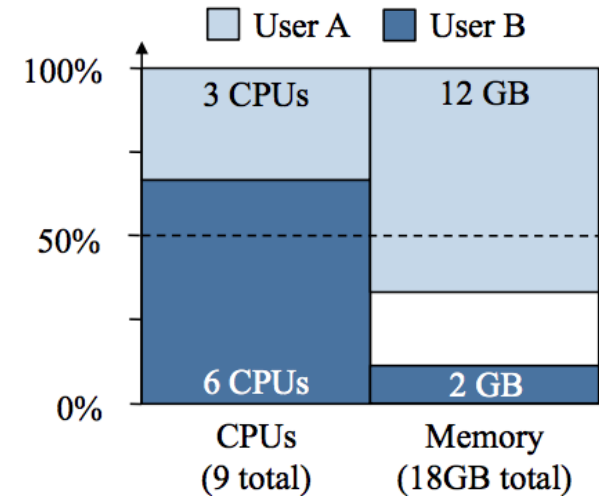
Equivalent to solving the welfare maximization for logarithms:  $\max_{x \in \mathcal{X}} \sum_{k=1}^K \log(x_k)$





# Multi-resource Fairness

- Two resources (CPU and memory)
- How to generalize fairness?



- Simple approach: “**Dominant Resource Fairness**”
  - Each user has a dominant resource share
  - Balance user shares with weighted MMF

A. Ghodsi et al., “Dominant Resource Fairness: Fair Allocation of Multiple Resource Types”, NSDI, 2011.

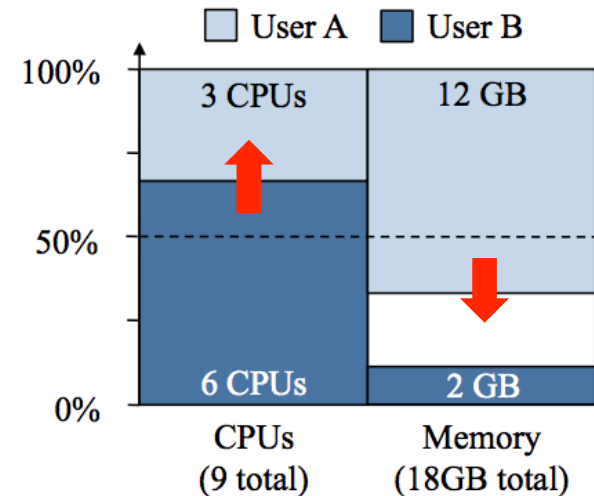
T. Bonald and J. Roberts, “Multi-Resource Fairness: Objectives, Algorithms and Performance”, ACM Sigmetrics, 2015.

# Dominant Resource Fairness

- Two users:
  - User A ( $w_{11}, w_{12}$ ) = (1,4)  
“4GBs for each CPU”
  - User B ( $w_{21}, w_{22}$ ) = (3,1)  
“1GB for each 3CPUs”

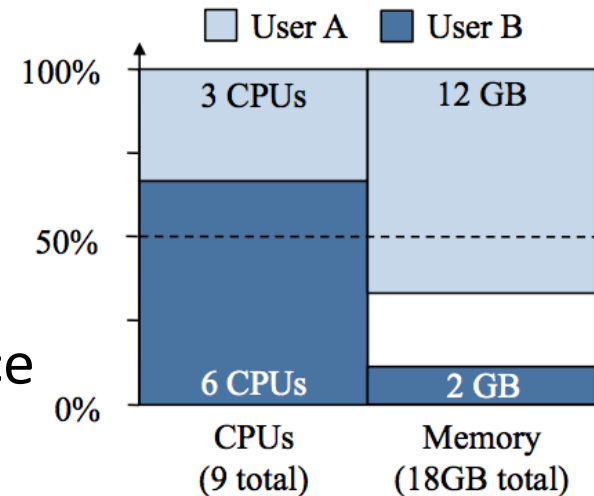
- Total of 9CPUs and 18GBs

- Dominant resource of user 1: 1/9, 4/18 -> memory
  - Dominant resource of user 2: 3/9, 1/18 -> CPU
- No concept of allocation yet



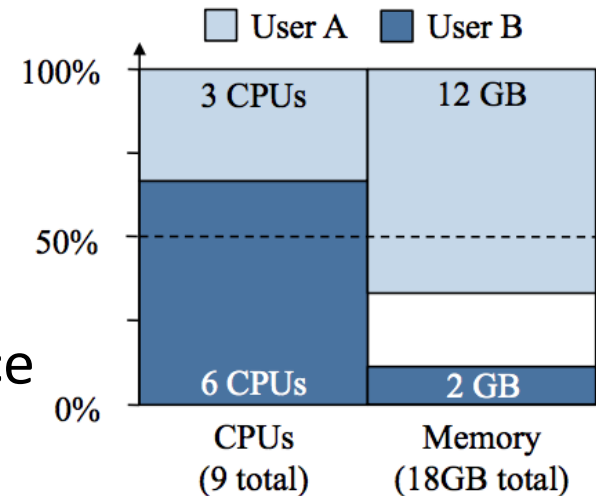
# Dominant Resource Fairness

- Two users:
  - User A ( $w_{11}, w_{12}$ ) = (1,4)
  - User B ( $w_{21}, w_{22}$ ) = (3,1)
- Change variables to dominant resource
  - $y_1 = 4/18 * x_1$
  - $y_2 = 3/9 * x_2$
- Quiz: solve for max-min fairness on Y



# Dominant Resource Fairness

- Two users:
  - User A ( $w_{11}, w_{12}$ ) = (1,4)
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- Change variables to dominant resource
  - $y_1 = 4/18 * x_1$
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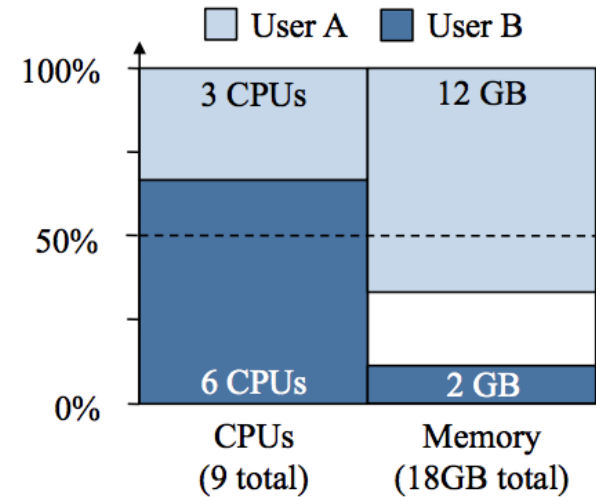
$$\begin{aligned} \text{for CPU, } & x_1 + 3x_2 \leq 9 \\ \text{for Memory, } & 4x_1 + x_2 \leq 18 \end{aligned}$$



$$\begin{aligned} \text{for CPU, } & y_1 + 2y_2 \leq 2 \\ \text{for Memory, } & 6y_1 + y_2 \leq 6 \end{aligned}$$

# Dominant Resource Fairness

- Two users:
  - User A ( $w_{11}, w_{12}$ ) = (1,4)
  - User B ( $w_{21}, w_{22}$ ) = (3,1)



- Solve:

$$\begin{aligned} \max_{y \geq 0} & \frac{y_1^{1-\alpha}}{1-\alpha} + \frac{y_2^{1-\alpha}}{1-\alpha} \\ \text{s.t.} & y_1 + 2y_2 \leq 2 \\ & 6y_1 + y_2 \leq 6 \end{aligned}$$



$$y_1 = y_2$$



$$y_1 = y_2 = 2/3$$



$$\begin{aligned} x_1 &= 18/4 * 2/3 = 3 \\ x_2 &= 9/3 * 2/3 = 6 \end{aligned}$$

User A: 3CPU, 12GB  
User B: 6CPU, 2GB

# Questions?

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- For questions about the course
- For questions about internship opportunities

<https://paschos.net/>

- Course material & relevant papers