

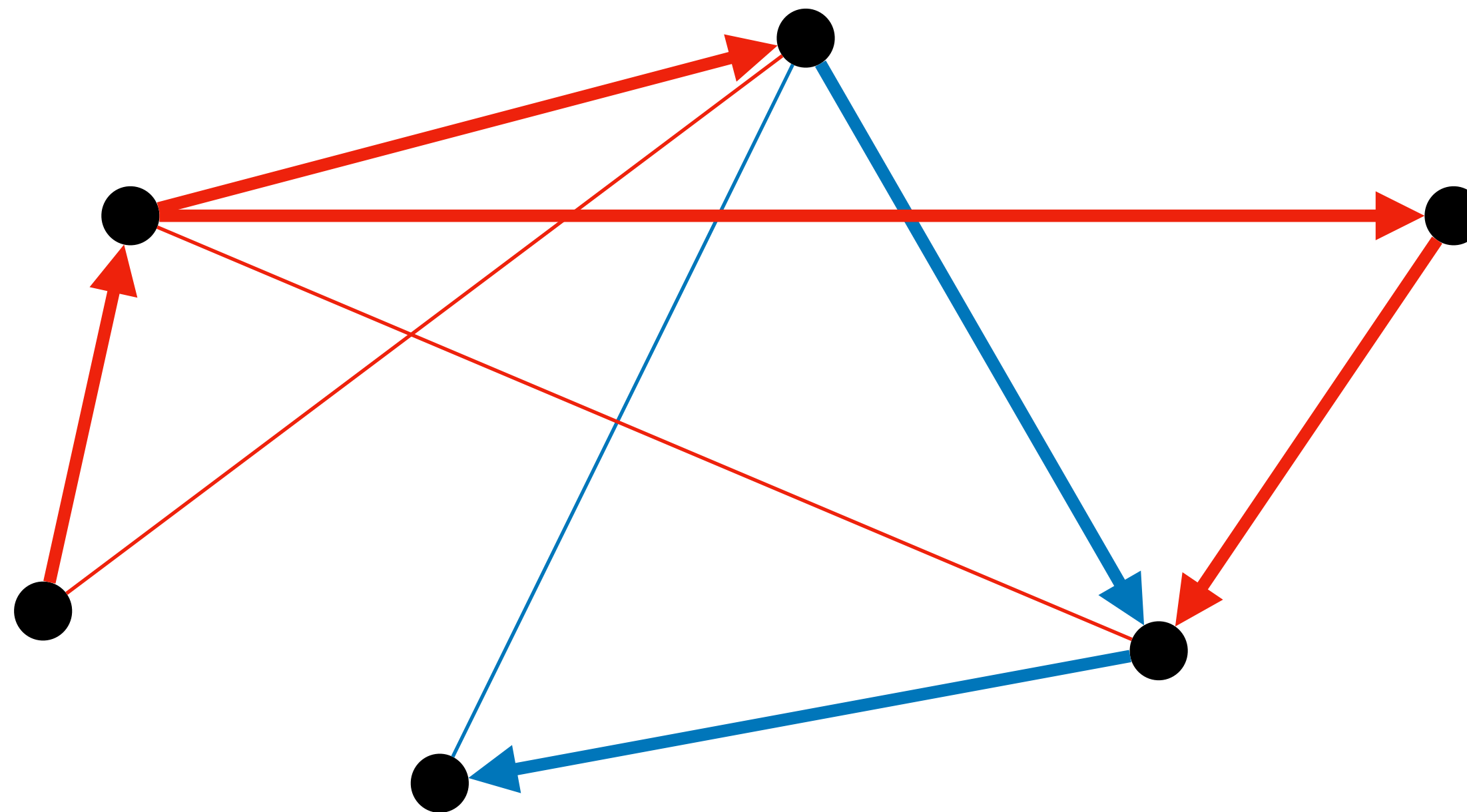
# Streaming CSPs

Santhoshini Velusamy (Toyota Technological Institute at Chicago)

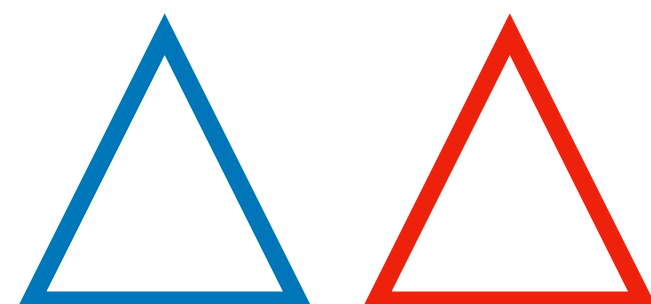
Sublinear Algorithms Program at the Simons Institute of Technology

# Constraint Satisfaction Problems (CSPs)

$n$  variables

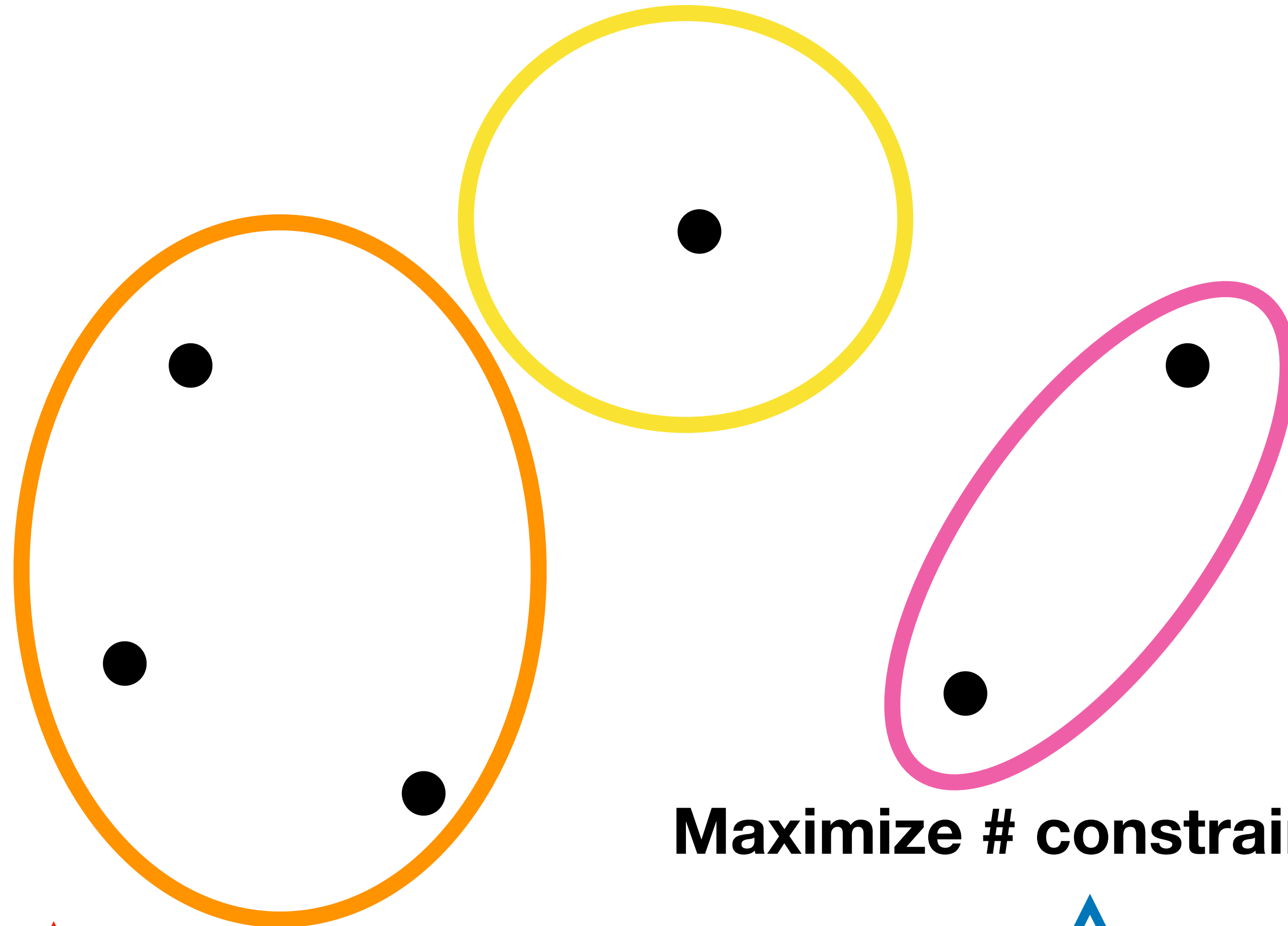


$k = 3$

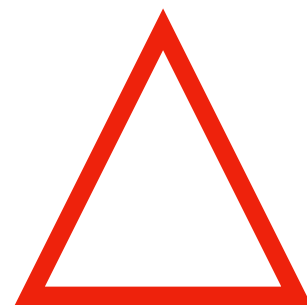


Constant number of constraint types of arity  $k$

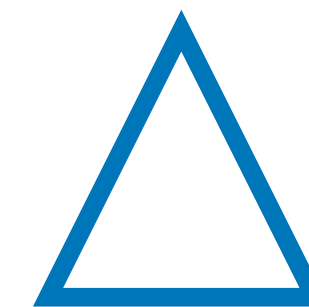
# Constraint Satisfaction Problems (CSPs)



**Maximize # constraints satisfied**

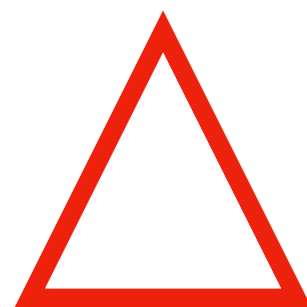
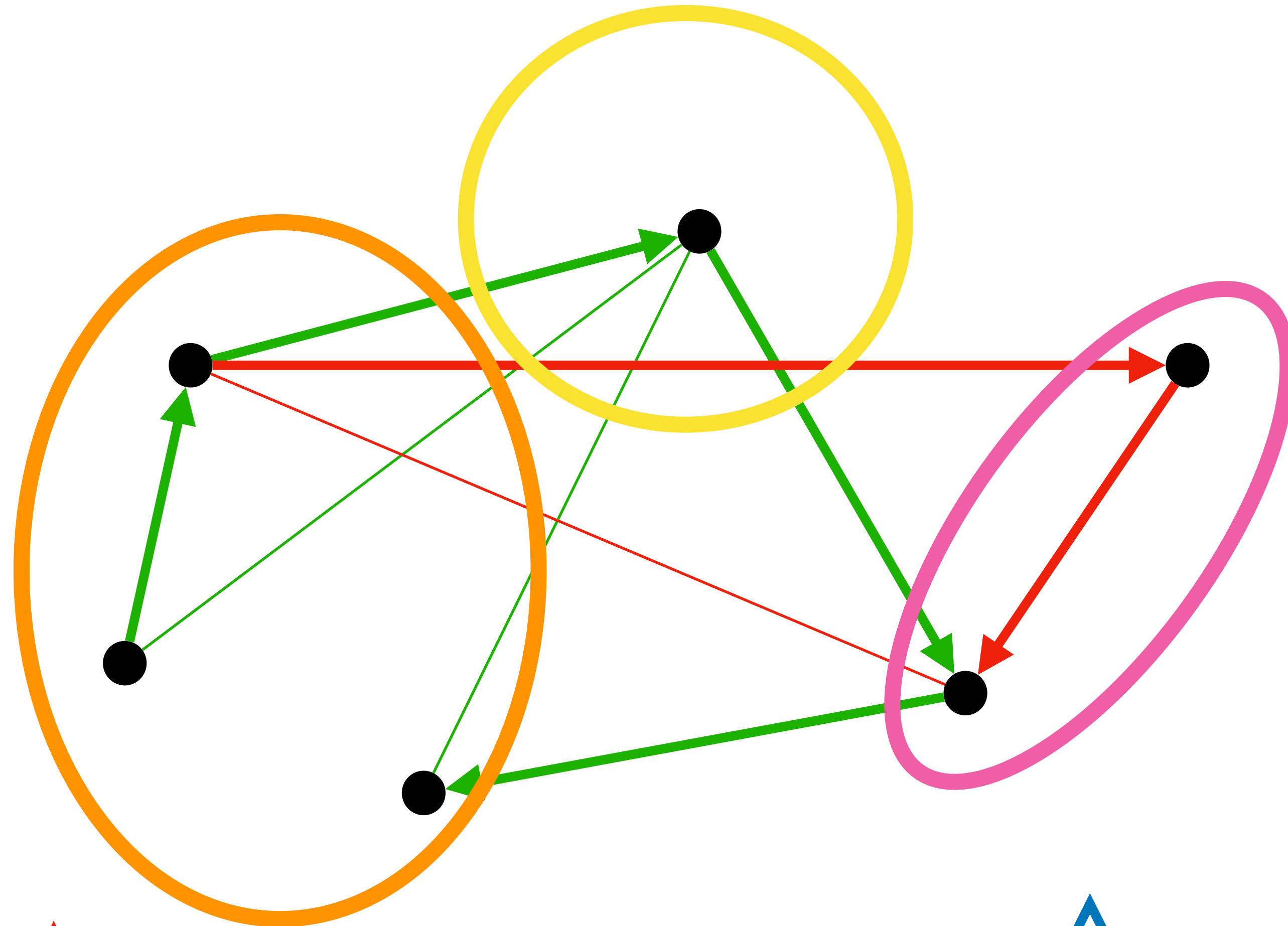


Last variable is **yellow**

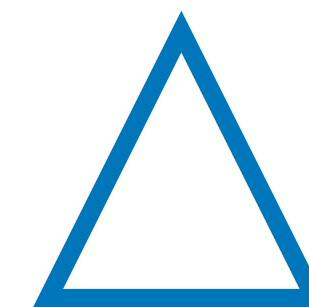


The variables are (**yellow**, **pink**, **orange**)

# Constraint Satisfaction Problems (CSPs)



Last variable is **yellow**

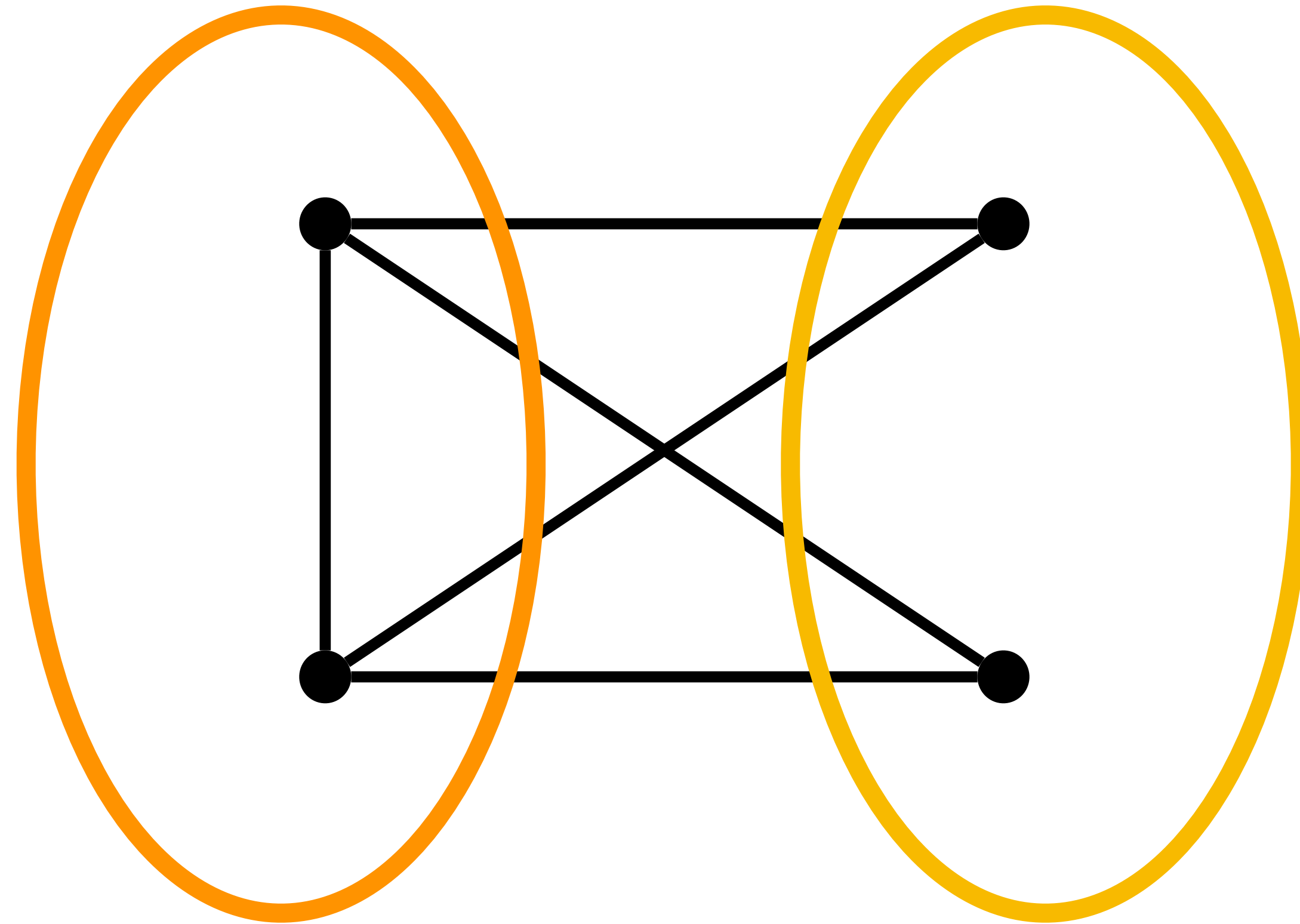


The variables are (**yellow**, **pink**, **orange**)

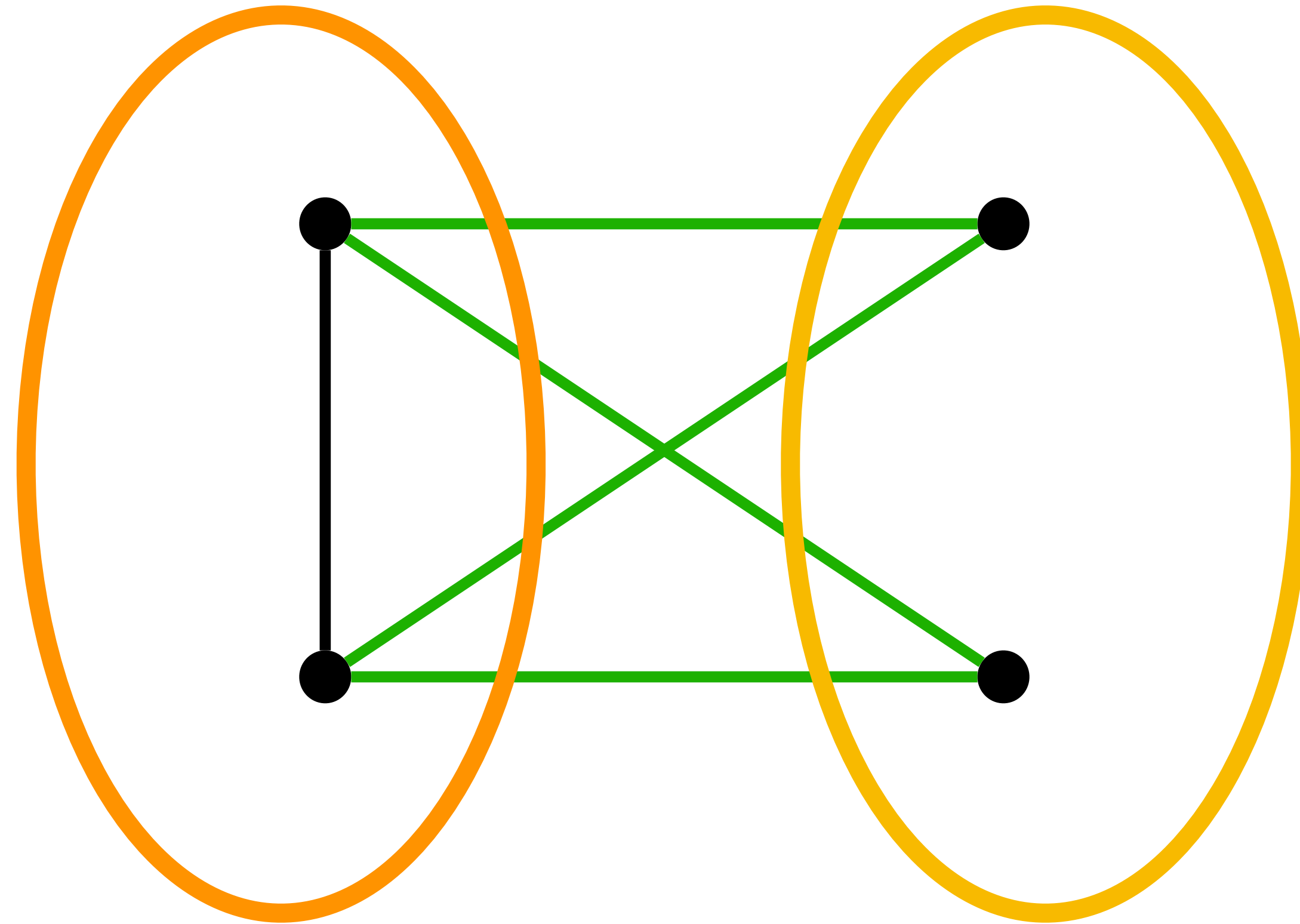
# Why study CSPs?

- Structure to study infinitely many problems simultaneously
- Contains several problems of interest: Max-CUT, Max-DICUT, Max- $k$ -SAT,...
- Allows for finite classifications through dichotomy theorems [[Schaefer '78](#), [Raghavendra '08](#), [Khot-Tulsiani-Worah '14](#), [Bulatov '17](#), [Zhuk '20](#), [Chou-Golovnev-Sudan-V '21](#), [Ghoshal-Lee '22](#), [Kol-Parmanov-Saxena-Yu '23](#)]
  - Insights about what lies at the heart of approximating CSPs
- Discovery of new techniques that are broadly applicable. For example, SDP rounding, Sum of squares, Unique Games,...

# Max-CUT



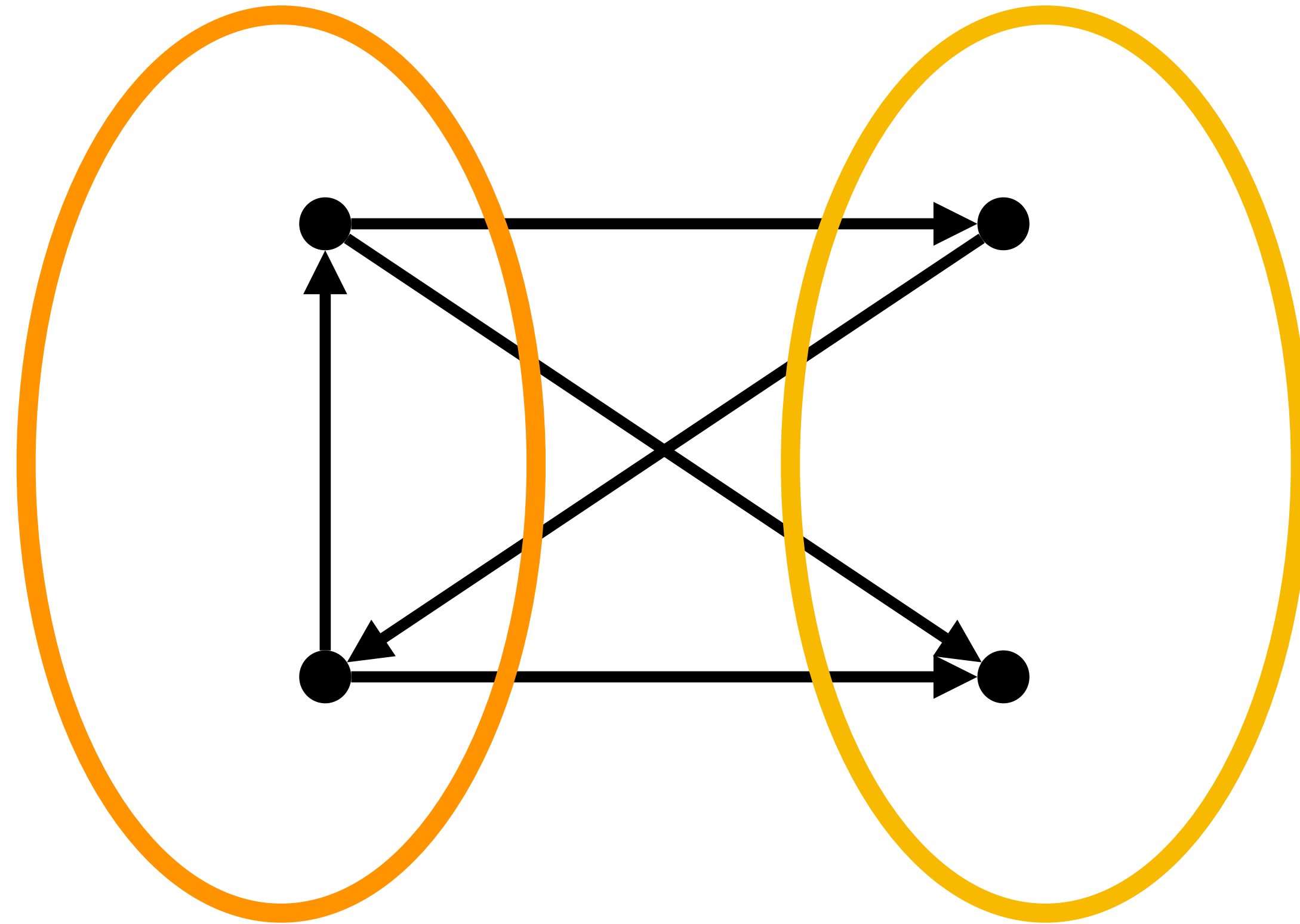
# Max-CUT



**CUT size = 4**

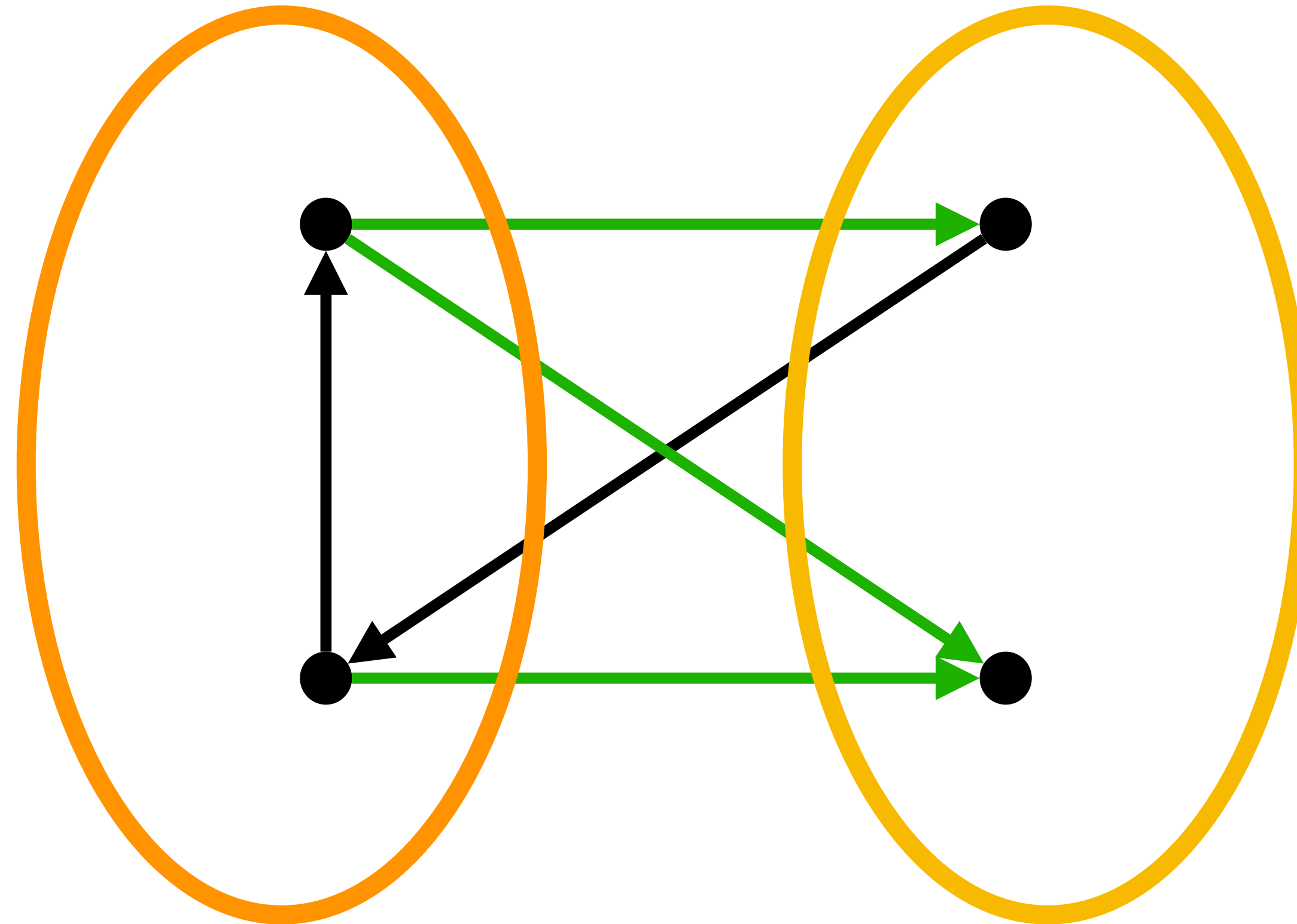
**Maximize CUT size**

# Max-DICUT





# Max-DICUT



**DICUT size = 3**

**Maximize DICUT size**

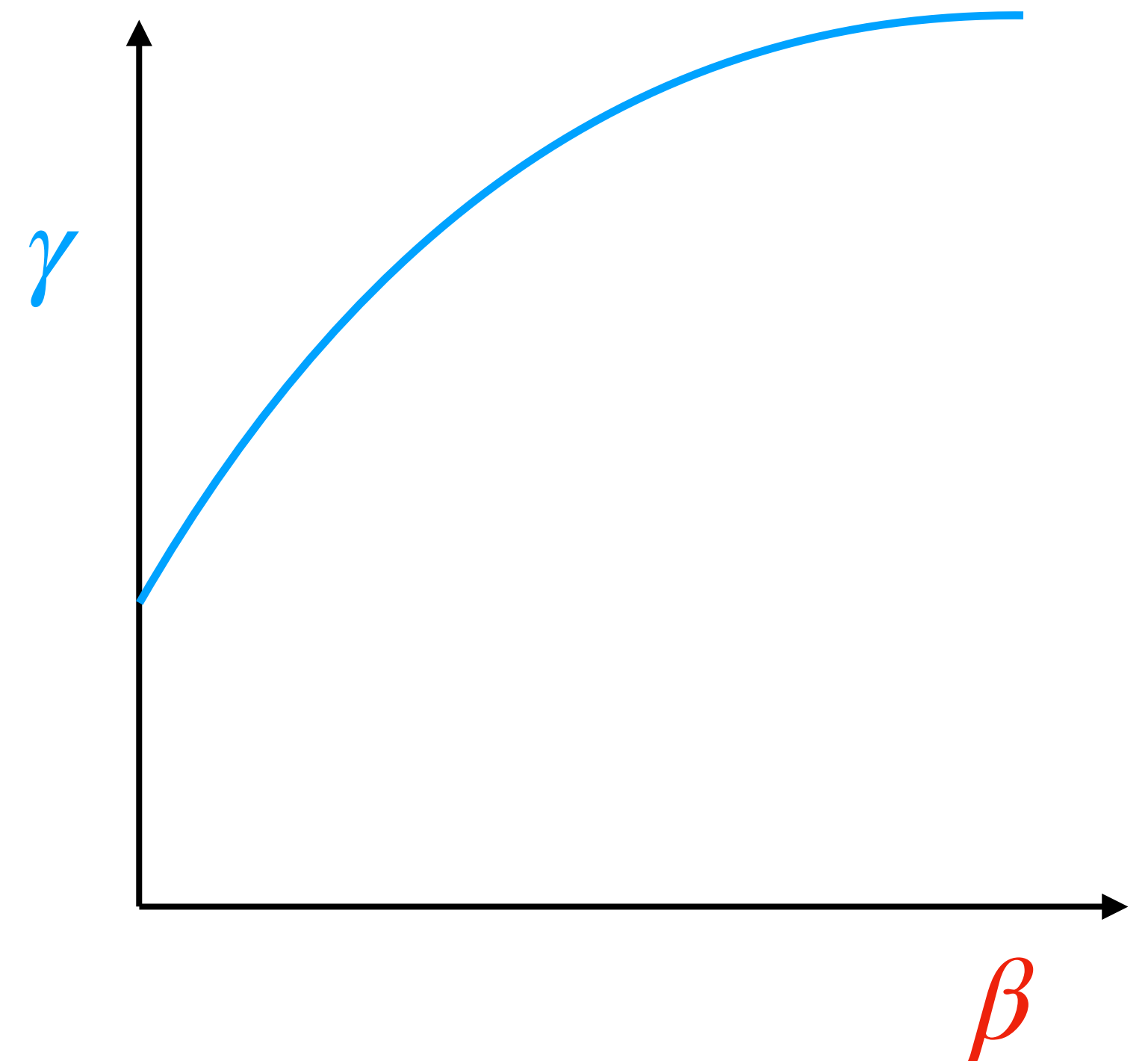
# Approximation algorithms

- Exact computation of optimum maybe NP-Hard!
- $0 < \alpha < 1$ ,  $\alpha$  approximation algorithm:
  - outputs  $T$  such that:  $\alpha \cdot \text{OPT} \leq T \leq \text{OPT}$
  - outputs an “underestimate” that is not off by a factor more than  $\alpha$
  - **Randomized algorithm**: outputs such an estimate with probability at least  $2/3$

# Promise problems

- For  $\beta < \gamma$ , can you distinguish instances with  $\text{OPT} \geq \gamma$  from  $\text{OPT} \leq \beta$ ?
- Finer study of approximation:

▶  $\alpha = \inf_{\text{indistinguishable } (\gamma, \beta)} \beta/\gamma$

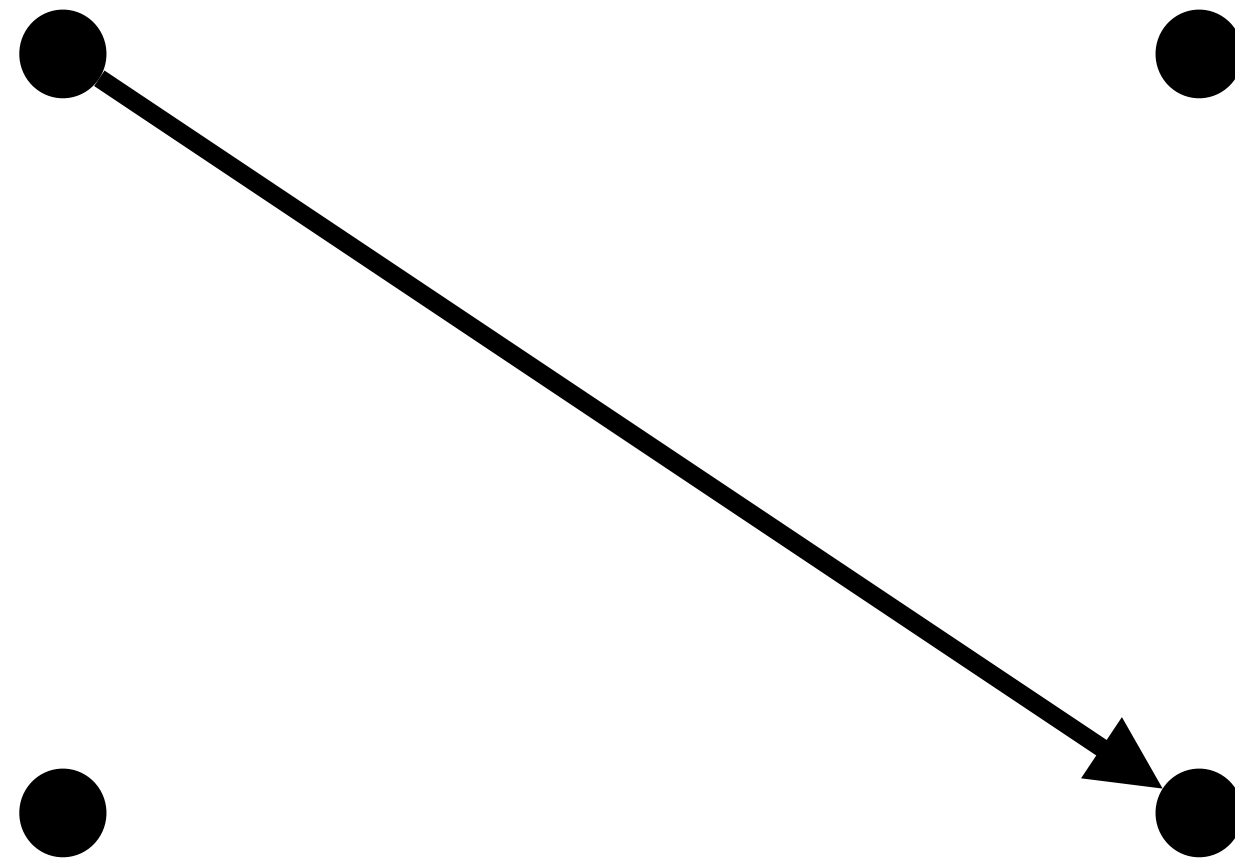


Approximability curve

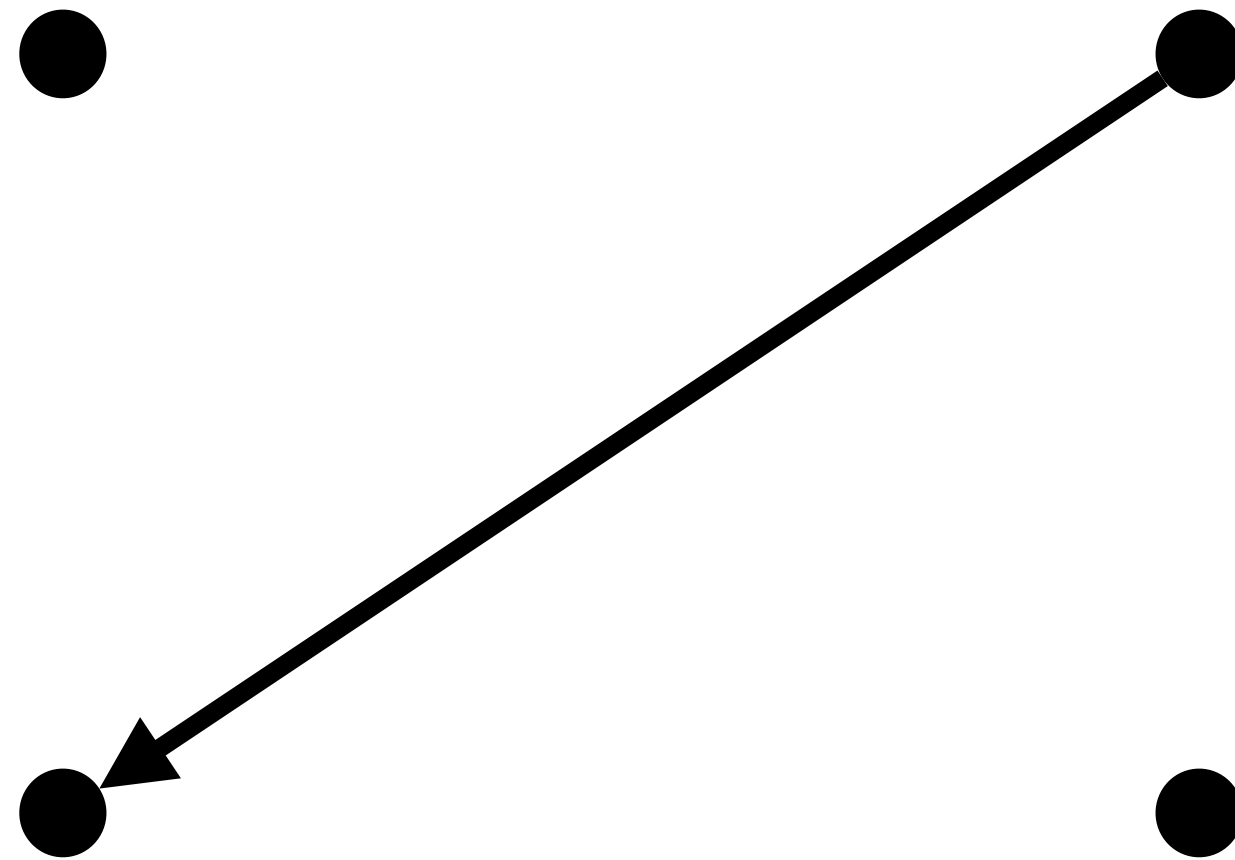
# Streaming setting



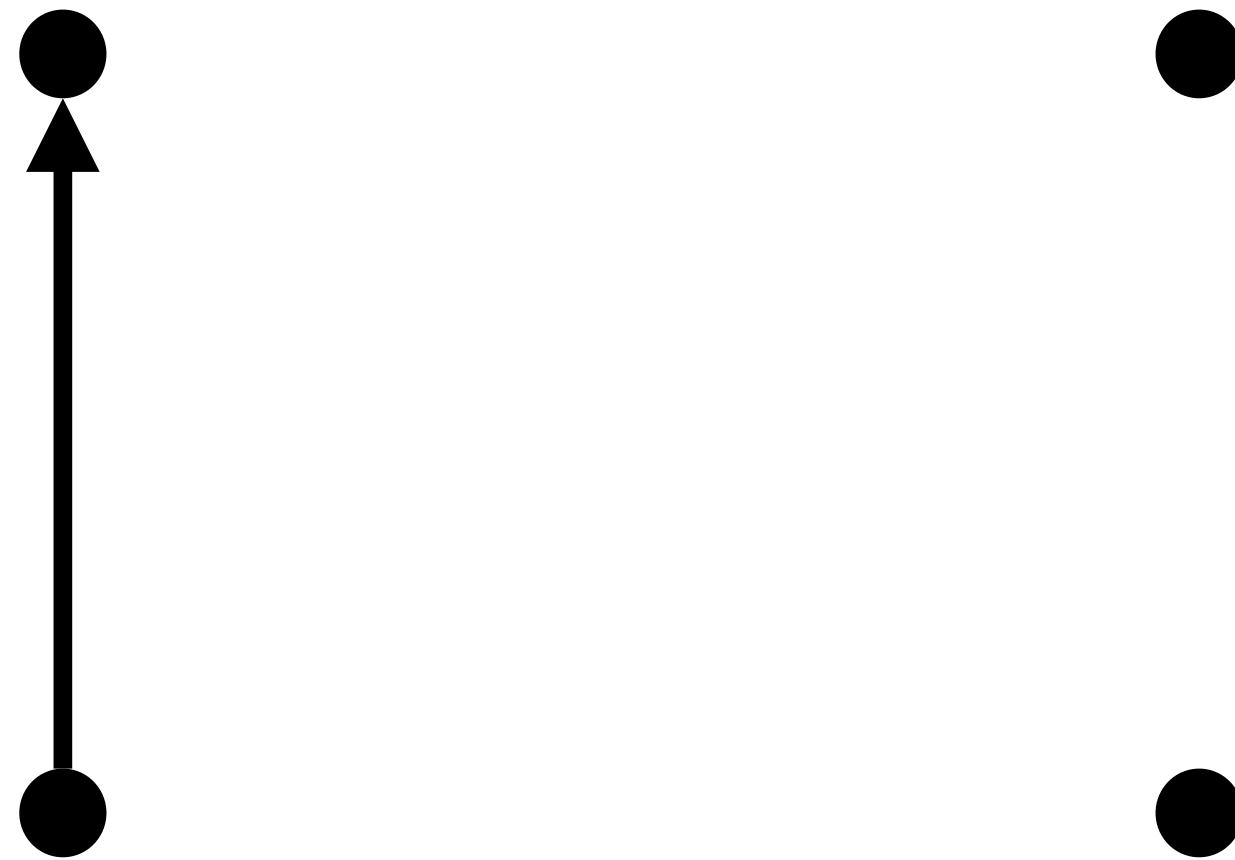
# Streaming setting



# Streaming setting



# Streaming setting



# Streaming setting



Compute the **size** of the largest CUT/DICUT in **small** space



# Folklore approximations

- Random CUT achieves  $1/2$  approximation for Max-CUT
  - [\[Kapralov-Krachun'19\]](#) Beating this requires at least  $\Omega(n)$  space!
- Random DICUT achieves  $1/4$  approximation for Max-DICUT
  - We can beat this in  $\log(n)$  space
- Sample  $O(n/\epsilon^2)$  random edges and compute Max-CUT/Max-DICUT value (possibly in exponential time) to obtain  $(1 - \epsilon)$  approximation!

# Streaming Approximability of CSPs in $o(n)$ space

# Streaming Approximability of CSPs in $o(n)$ space

(bounded-degree instances)

# Dichotomy theorem: $\log n$ vs $\sqrt{n}$

[Chou-Golovnev-Sudan-V'21]

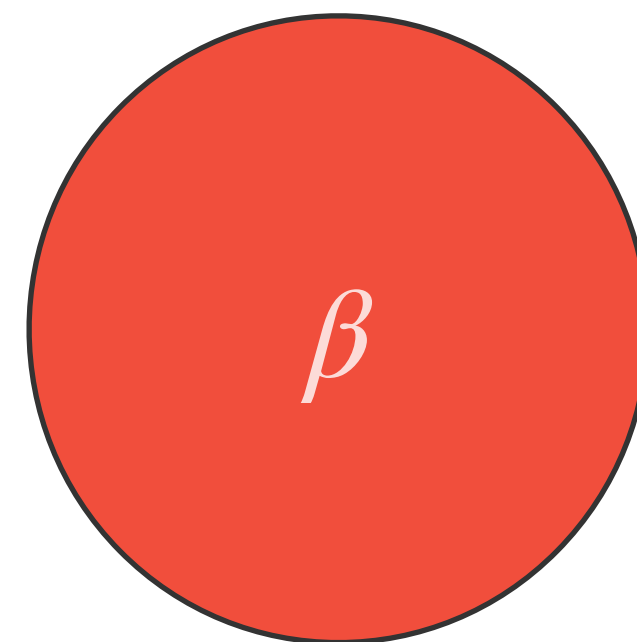
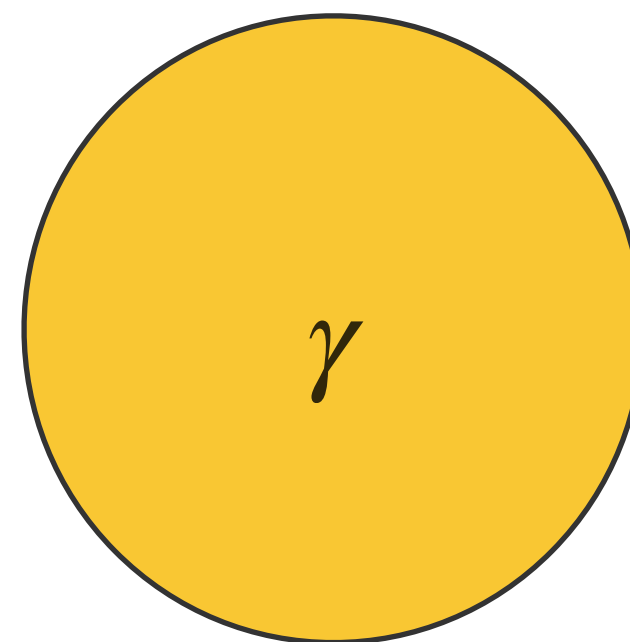
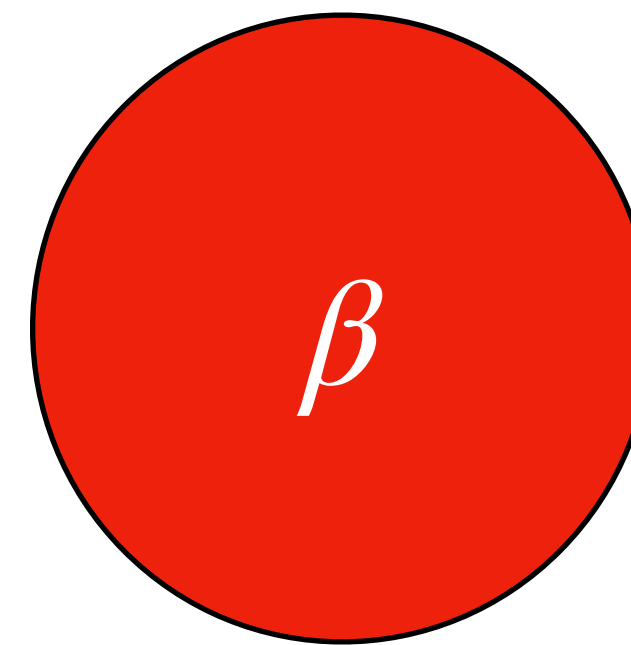
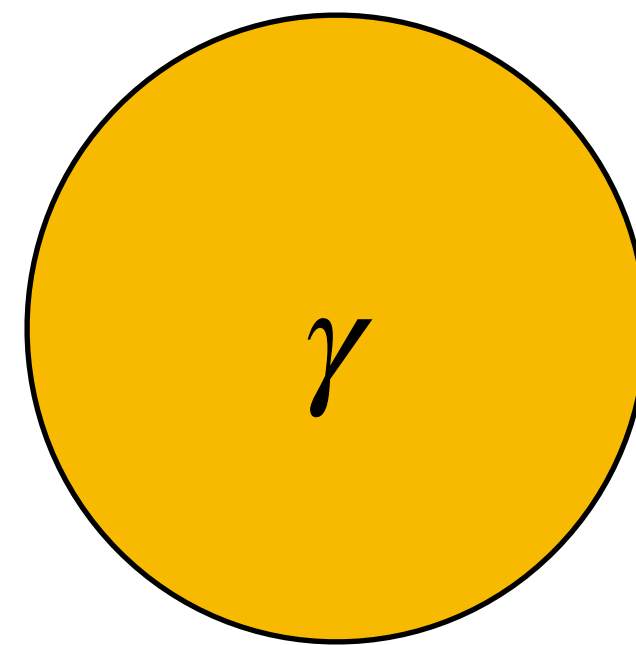
- Bias of a vertex  $v = \frac{\text{out}(v) - \text{in}(v)}{\text{deg}(v)} \in [-1, 1]$
- Bias intervals:  $-1 \leq b_0 \leq \dots \leq b_{r-1} \leq 1$  where  $b_{i+1}/b_i \leq \epsilon$
- Vertex-bias statistics: number of vertices of degree  $i \in [d]$  in each bias class
- Can be computed in  $O(\log n)$  space

# Dichotomy theorem: $\log n$ vs $\sqrt{n}$

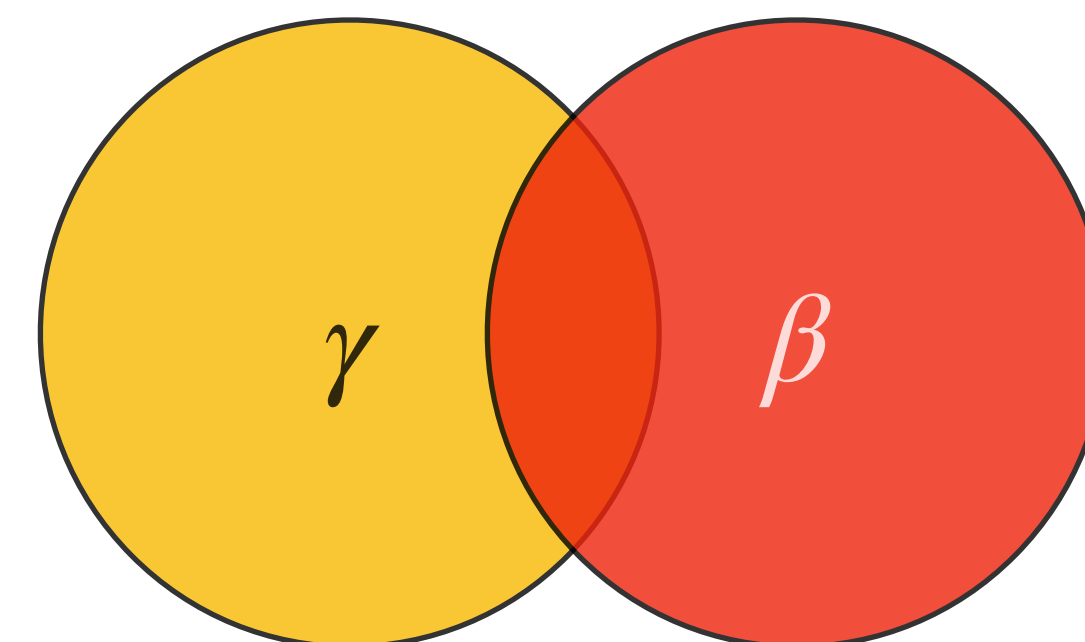
[Chou-Golovnev-Sudan-V'21]

For  $\beta < \gamma$ , when can you distinguish instances with  $\text{OPT} \geq \gamma$  from  $\text{OPT} \leq \beta$ ?

vertex-bias statistics



distinguishable in  $O(\log n)$  space



indistinguishable in  $o(\sqrt{n})$  space

# Dichotomy theorem: $\log n$ vs $\sqrt{n}$

[Chou-Golovnev-Sudan-V'21]

- Decidable in PSPACE
- Decidable characterization of CSPs that are “approximation resistant” in  $o(\sqrt{n})$  space
- **Incomplete**: only proves  $\sqrt{n}$  space lower bounds
- Linear space lower bounds for an infinite subset of approximation resistant CSPs [Chou-Golovnev-Sudan-Velingker-V'22]

# $\tilde{O}(\sqrt{n})$ space algorithm for Max-DICUT

[Saxena-Singer—Sudan-V'23]

- Edge-bias statistics
- Compute the number of edges between each pair of bias classes
- Sample  $O_d(\sqrt{n})$  vertices and record their induced subgraph
- Oblivious algorithms: “round” all variables in each bias class with same probability [Feige-Jozeph'15]
- Better approximation-ratio for Max-DICUT!
- $\exists \delta > 0$  such that oblivious algorithm cannot beat  $1/2 - \delta$  approximation

# Reaching $1/2$ -approximation

[Saxena-Singer—Sudan-V'upcoming]

- Radius- $k$ -subgraph statistics  $\implies 1/2 - 1/k^2$  approximation
- Computable in  $\tilde{O}(n^{1-f(d,k)})$  space
- Sample  $O(n^{1-f(d,k)})$  vertices and record their induced subgraph



**Open problems**

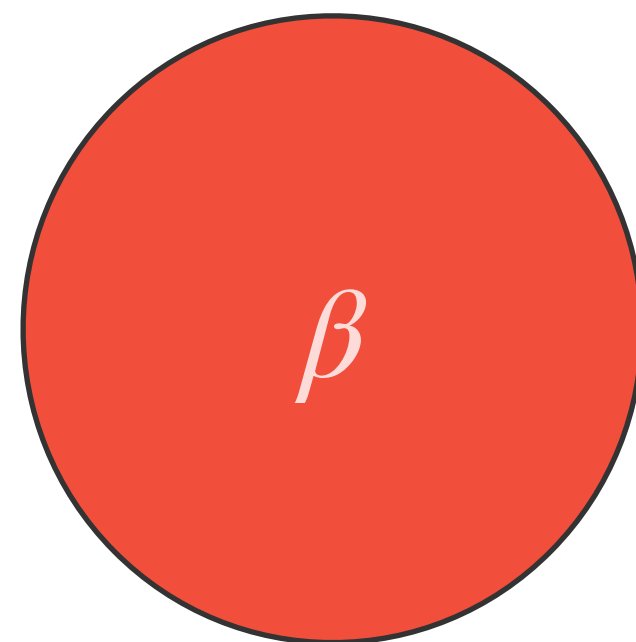
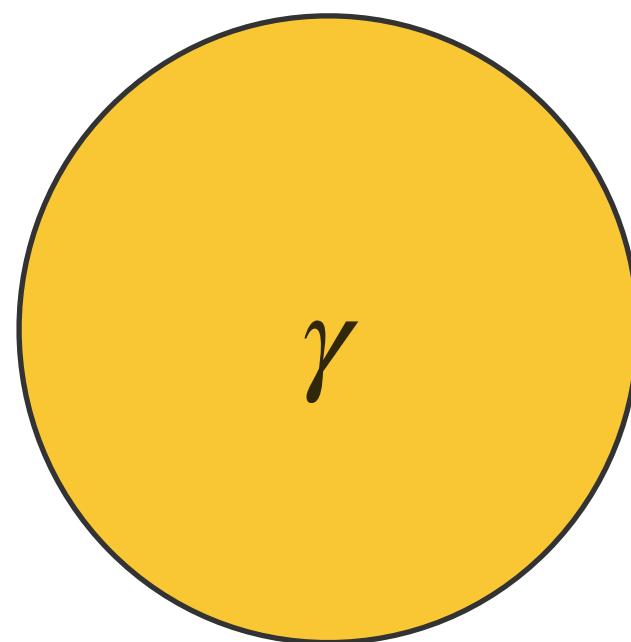
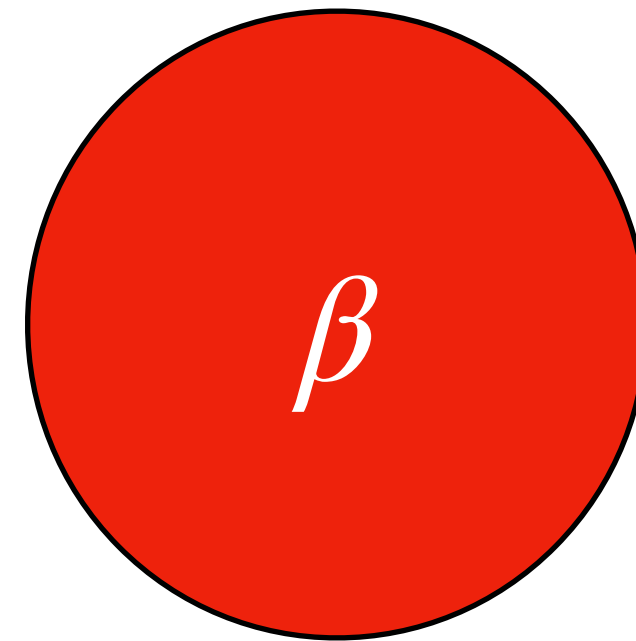
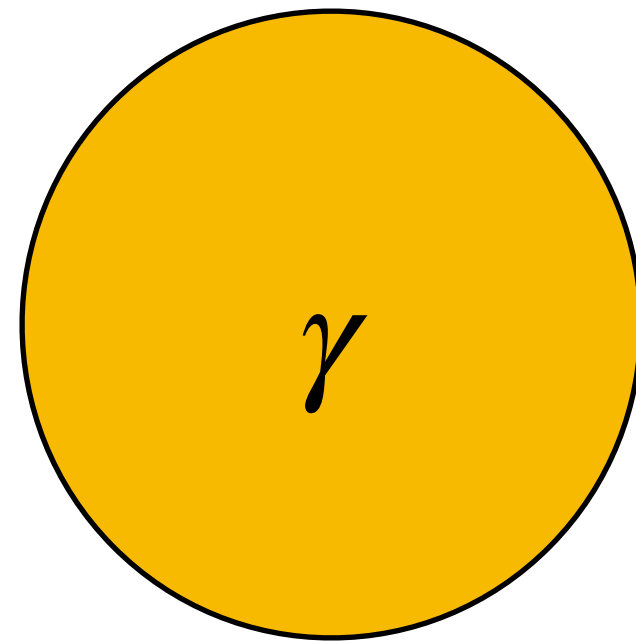
**Do edge-bias statistics give the best algorithm for Max-DICUT in  $\tilde{O}(\sqrt{n})$  space?**

**Do edge-bias statistics give the best  
algorithm for Max-CSP in  $\tilde{O}(\sqrt{n})$   
space?**

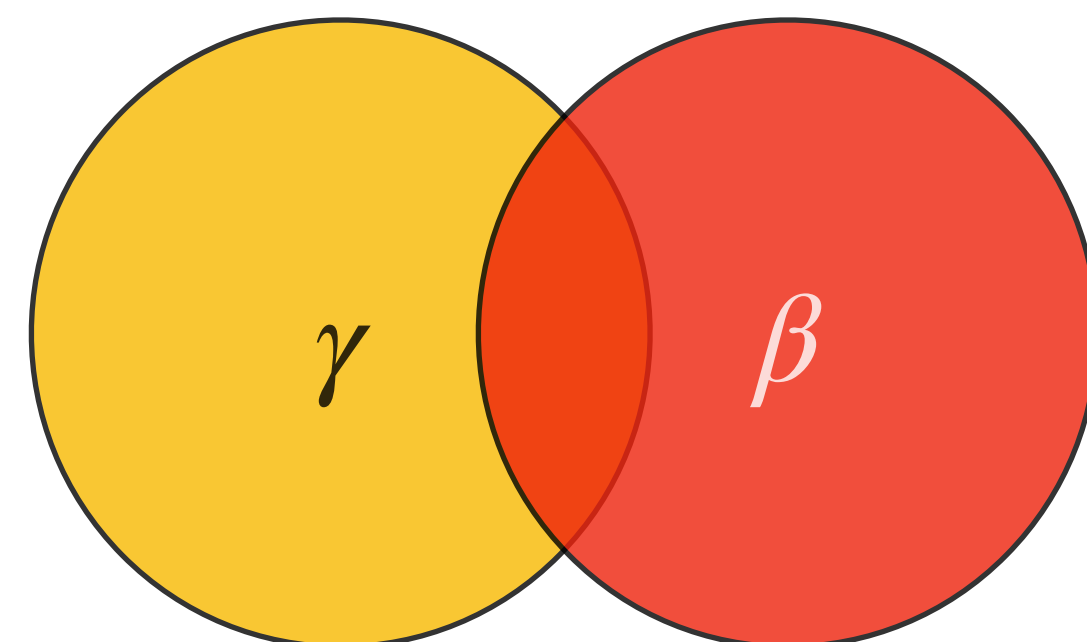
# Dichotomy theorem??

For  $\beta < \gamma$ , when can you distinguish instances with  $\text{OPT} \geq \gamma$  from  $\text{OPT} \leq \beta$ ?

edge-bias statistics



distinguishable in  $\tilde{O}(\sqrt{n})$  space



indistinguishable in  $o(n^c)$  space,  
 $c > 1/2$

**Is there a CSP that is approximation  
resistant in  $o(\sqrt{n})$  space but  
approximable in  $o(n)$  space?**

**Is every CSP that is approximation  
resistant in P also approximation  
resistant in  $o(\sqrt{n})$  space?**

**Thank you!**