

# Hardness of Approximation meets Parameterized Complexity

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# Global Outline

Day 1: The Setting

Day 2: Gap Creation

Day 3: Applications

# Day 1 Outline

## Part 1: Hardness of Approximation

- Hardness of Approximation in NP
- Hardness of Approximation in Parameterized Complexity

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## Part 2: Key Problems in Parameterized Inapproximability

- MaxCover
- One-Sided Biclique

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- Hardness of Approximation in Parameterized Complexity

## Part 2: Key Problems in Parameterized Inapproximability

- MaxCover
- One-Sided Biclique

## Part 3: Coding Theory

- Definition and Geometric Intuition
- Random Codes
- Algebraic Codes

# Part 1

## Hardness of Approximation

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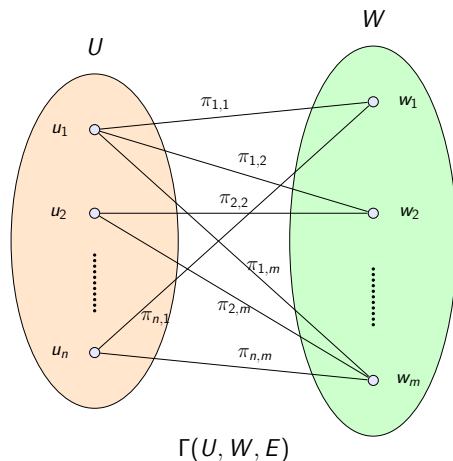
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PCP Theorem: Bedrock of  
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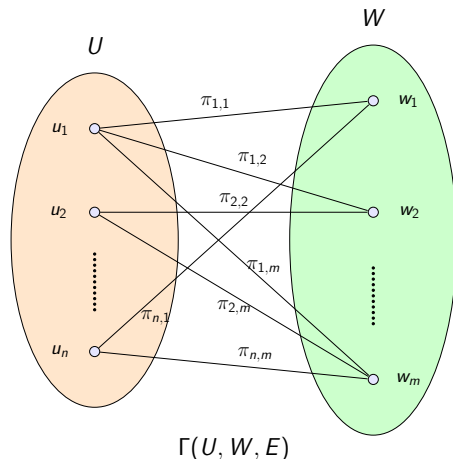
# PCP Theorem & Label Cover



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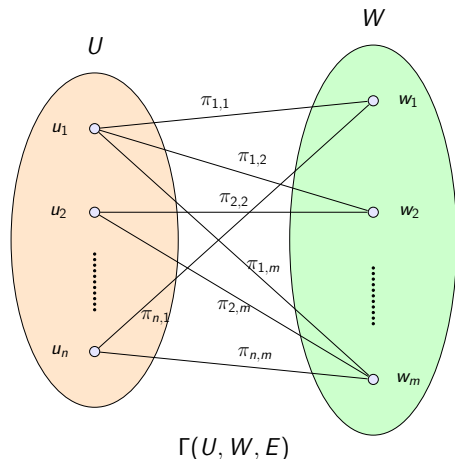


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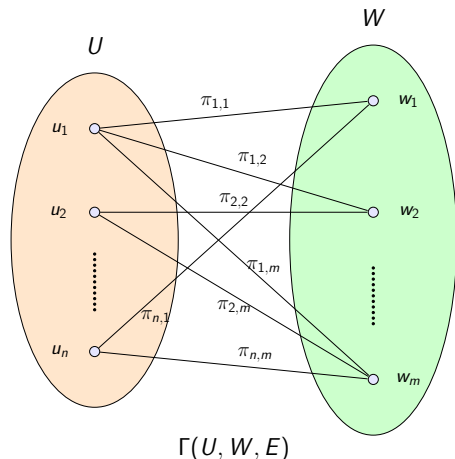
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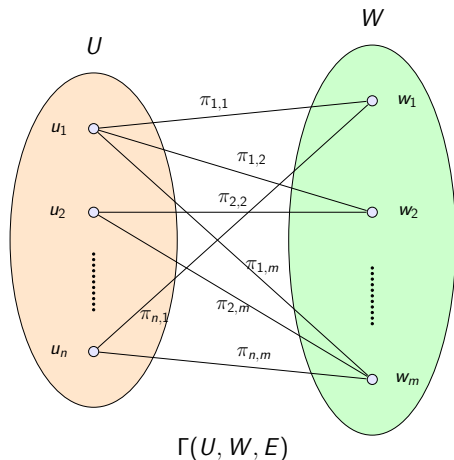
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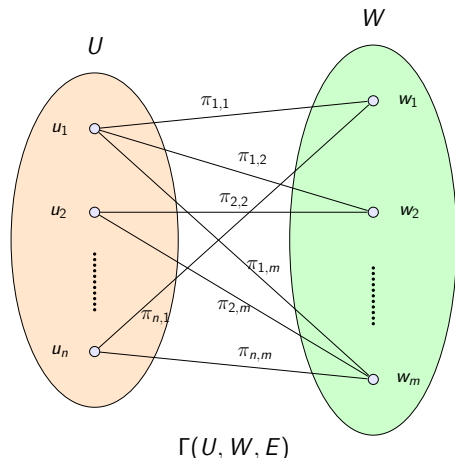
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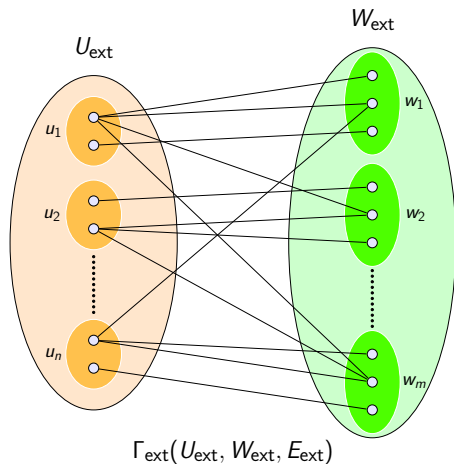
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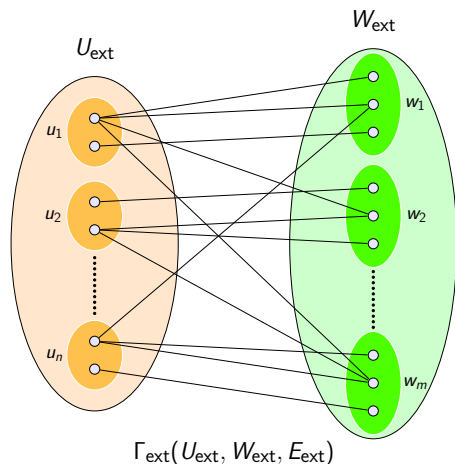
$$\text{VAL}(\Gamma) = \max_{\sigma_U, \sigma_W} \text{VAL}(\Gamma, \sigma_U, \sigma_W)$$

Determining if  $\text{VAL}(\Gamma) = 1$  or  
if  $\text{VAL}(\Gamma) \leq 0.99$  is NP-Hard

# Extended Label Cover



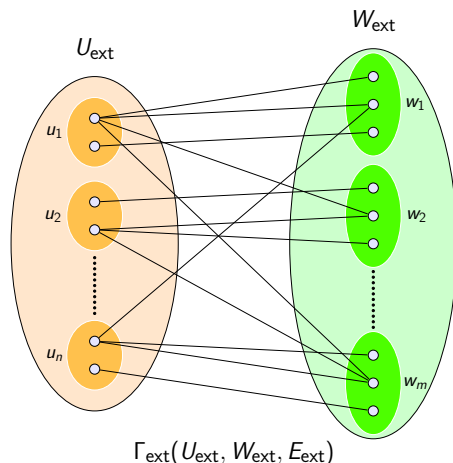
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$n \cdot |\Sigma_U|$  nodes in  $U$   
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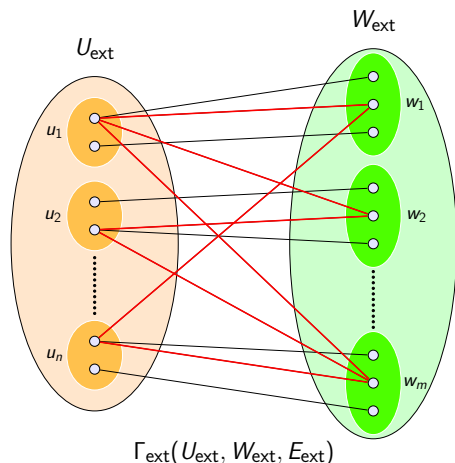
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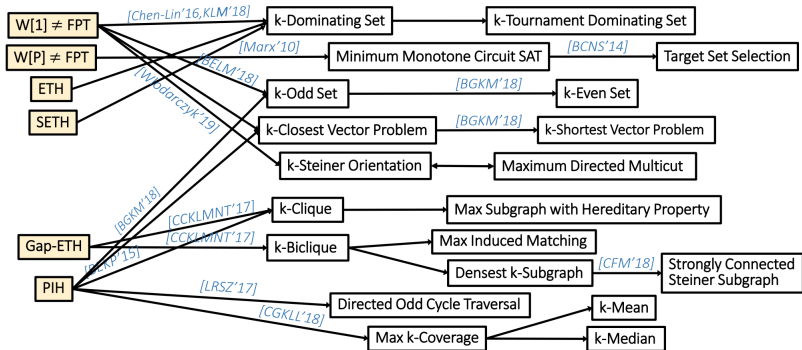
# Parameterized Inapproximability: Motivation

- Many Optimization problems are NP-Hard
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- Set Cover: Hard to cope!
- New direction: Fixed Parameter Approximability

Is there a  $F(k) \cdot \text{poly}(n)$  time algorithm that approximates to a factor  $T(k)$ ?

# Parameterized Inapproximability: Partial Summary

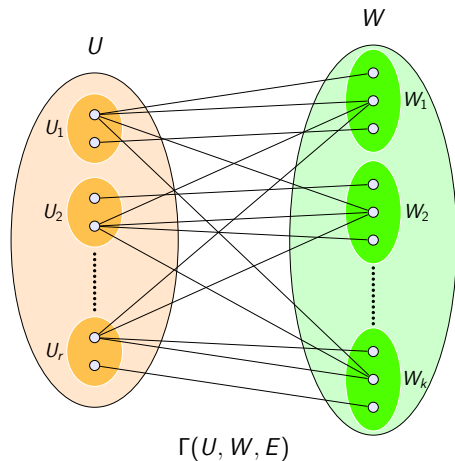
## Parameterized Inapproximability: *Recent Developments*



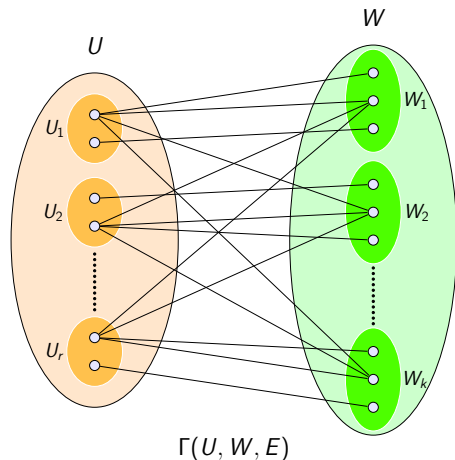
# Part 2

## Key Problems in Parameterized Inapproximability

# MaxCover [Chalermsook et al. 2017]

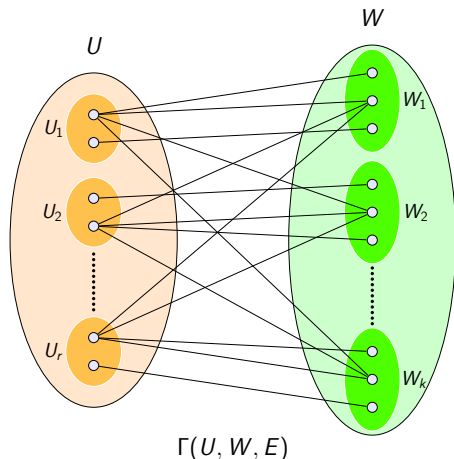


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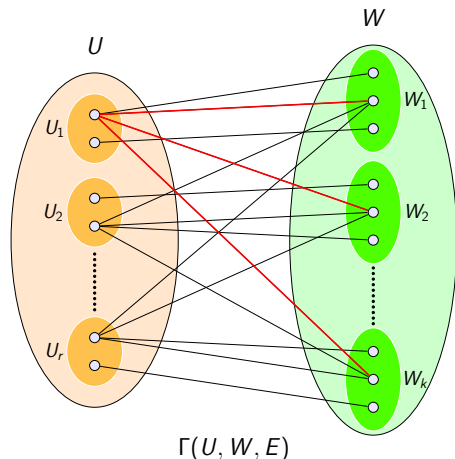
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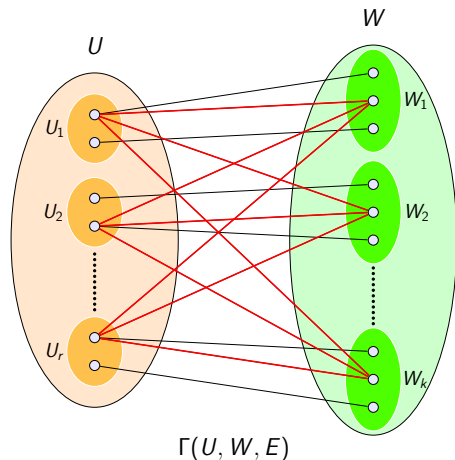


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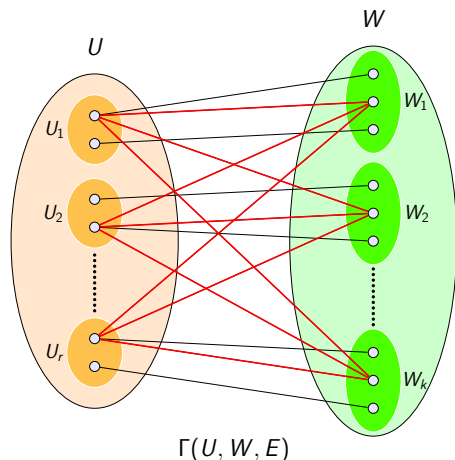
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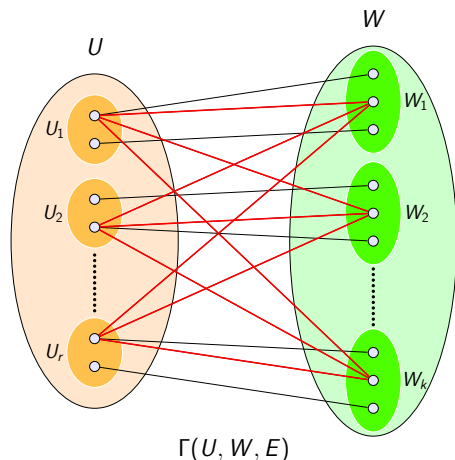
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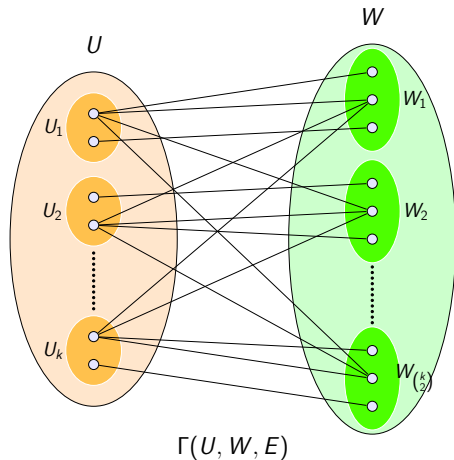
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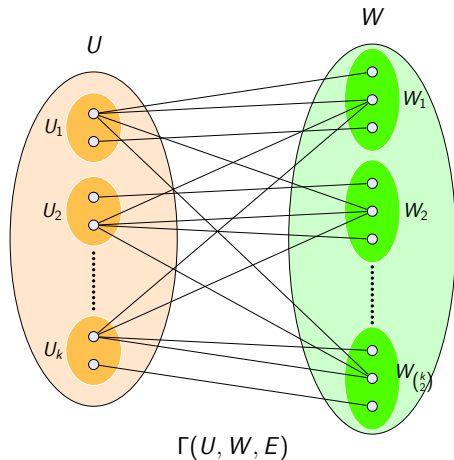
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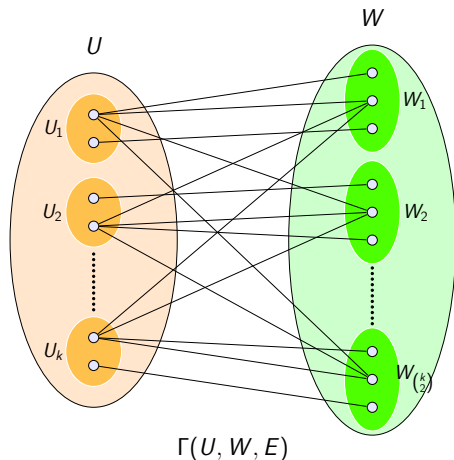


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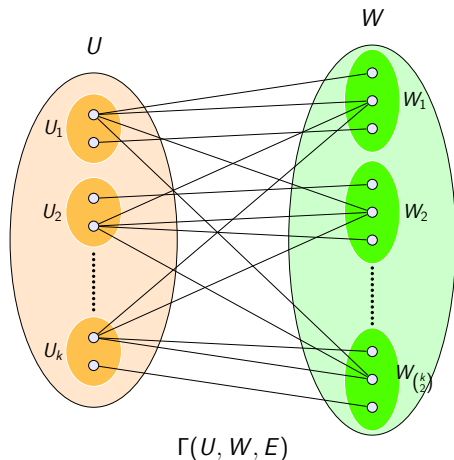
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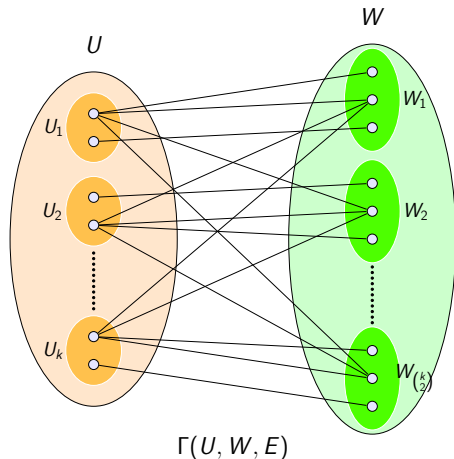


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# MaxCover: Results

- W[1]-Complete if there are  $F(k)$  left super nodes



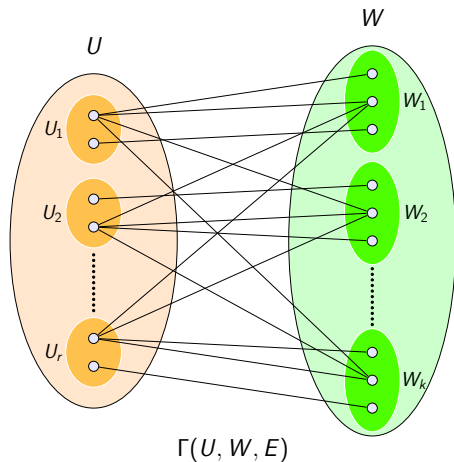
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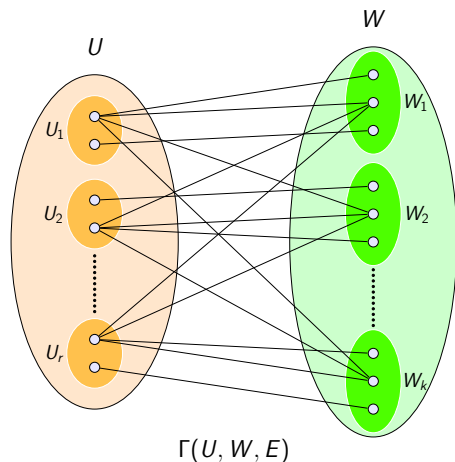
# MaxCover: Results

- W[1]-Complete if there are  $F(k)$  left super nodes
- 1 vs.  $k/n^{1/\sqrt{k}}$  is W[1]-Hard
- Central problem to understand parameterized inapproximability of Set Cover and Clique

# MaxCover: $W[1]$ Membership

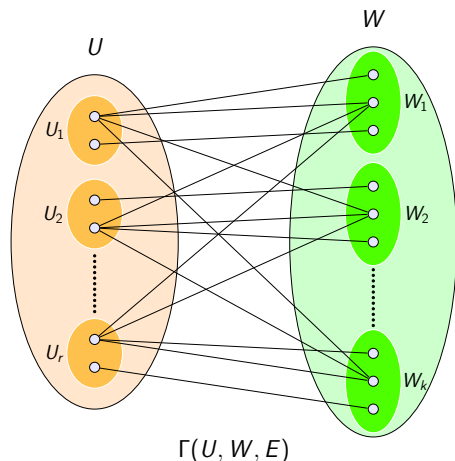


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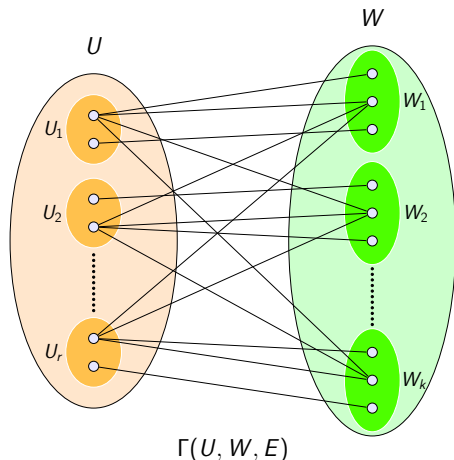
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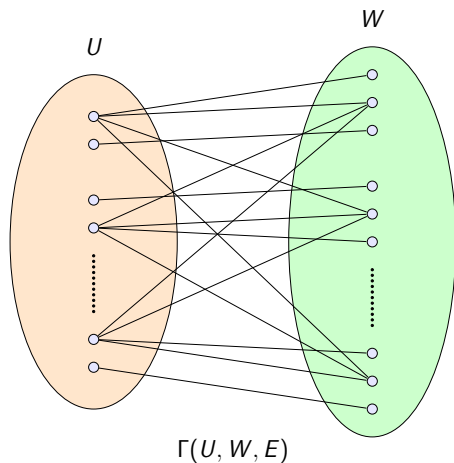


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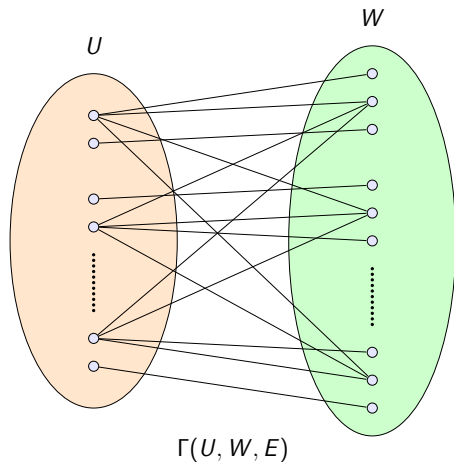
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MaxCover from **ETH** and **SETH**  
have  $r = F(k)$

# One-Sided Biclique



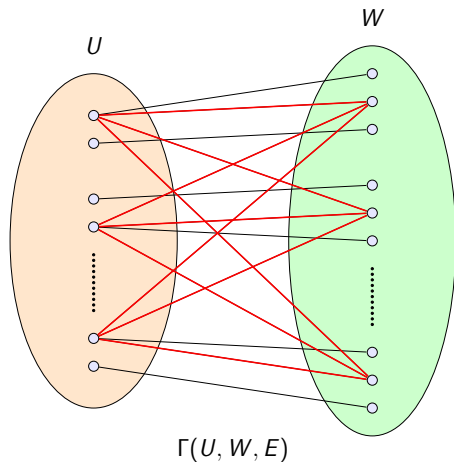
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# One-Sided Biclique vs. MaxCover

- Colored vs. Non-colored

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# One-Sided Biclique vs. MaxCover

- Colored vs. Non-colored
- Covering vs. Common neighbors
- One-Sided Biclique reduces to MaxCover: Color Coding
  - What about the other direction?

# Summary

- Hardness of Approximation meets Parameterized Complexity: New **Exciting** Area!
- MaxCover and One-Sided Biclique are **key** problems for which we have proved inapproximability results.

# Part 3

## Coding Theory

# Coding Theory: Geometric Motivation

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# Coding Theory: Definitions

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- Distance of  $C$ :

$$\Delta(C) := \min_{x, y \in C} \|x - y\|_0$$

A **good** code: for  $\rho, \delta > 0$ ,  $|C| = 2^{\rho L}$ ,  $\Delta(C) = \delta L$ .

# Random Codes

## Random Strings are Good Codes

For some small  $\rho > 0$ , if we pick  $2^{\rho L}$  random strings uniformly and independently then they form a code with distance at least  $1/4$  (whp).



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Many Efficient Deterministic Good Codes Exist!

# Coding Theory: Reed Solomon Codes

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- Reed Solomon Codes meet the Singleton bound!

# Tomorrow's plan

- **MaxCover**: Gap Creation by using Codes
- **One-Sided Biclique**: Gap creation by using Random Graphs/Polynomials