



A New  
Direction

JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

# A New Direction: The Oriented Chromatic Number of Random Graphs of Bounded Degree

JD Nir

University of Manitoba



Toronto Metropolitan University

June 1, 2022



# Motivation

## Random Graphs

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### Definition

A **random graph**  $\mathcal{G}$  is a probability distribution over graphs, though we often think of it as a random process that results in a graph.



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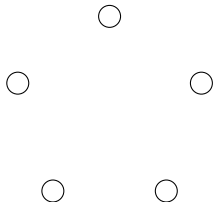
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Some examples:

- $\mathcal{G}_{n,p}$ :  $n$  vertices, each possible edge included with probability  $p$ .





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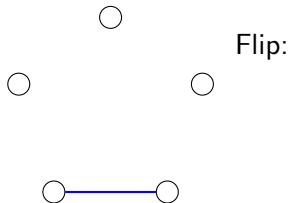
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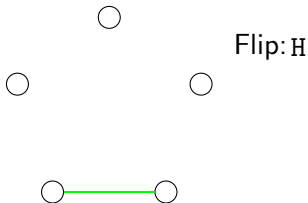
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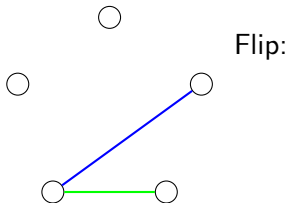
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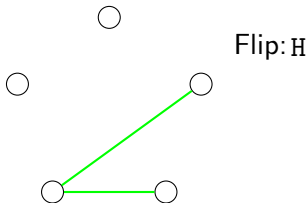
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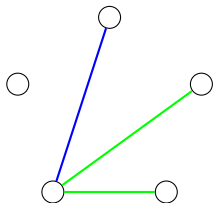
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Flip:





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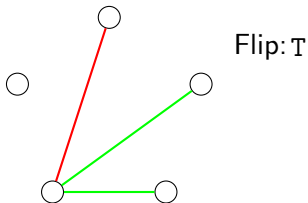
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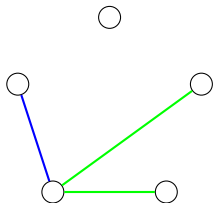
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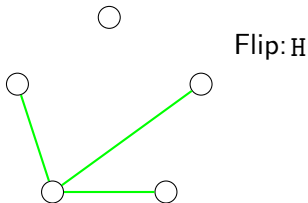
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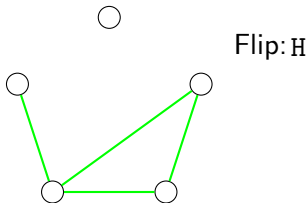
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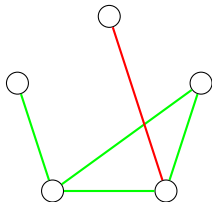
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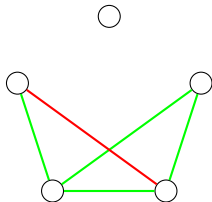
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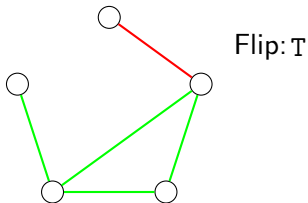
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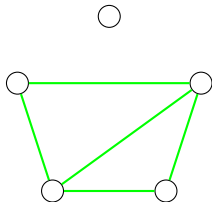
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Flip:  $\mathbb{H}$





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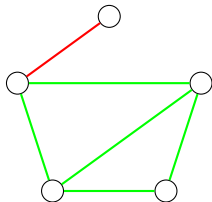
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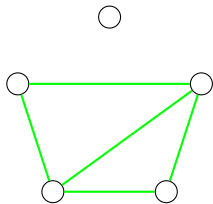
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- $\mathcal{G}_{n,p}$ :  $n$  vertices, each pair adjacent with probability  $p$ .
- $\mathcal{G}_{n,d}$ :  $d$ -regular graph of order  $n$ , selected uniformly.



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Different random graph models produce different distributions.



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## Graph Colouring

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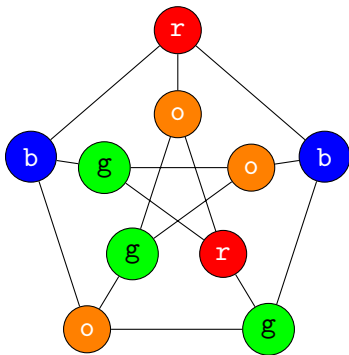
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### Definition

A **vertex colouring** of a graph is an assignment of colours to vertices such that adjacent vertices receive different colours.





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## Graph Colouring

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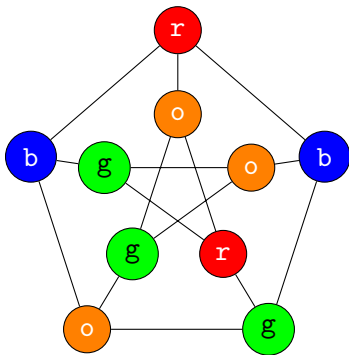
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### Definition

A **vertex colouring** of a graph is an assignment of colours to vertices such that adjacent vertices receive different colours. The **chromatic number** of a graph  $G$  is the smallest  $k$  such that  $G$  admits a vertex colouring with  $k$  colors.





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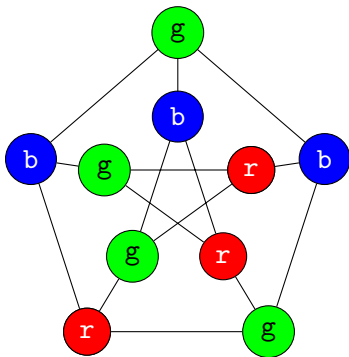
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# Motivation

## Chromatic Number of Random Graphs

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### Definition

The chromatic number of a random graph  $\chi(\mathcal{G})$  is a random variable that takes value  $k$  with the probability that  $G \sim \mathcal{G}$  has  $\chi(G) = k$ .



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Example: if  $\mathcal{G}$  is the collection of cycles of length  $3 \leq \ell \leq 100$  with uniform probability then

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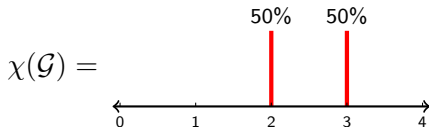
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Asymptotic certainty

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## Question

For  $p = \frac{1}{2}$ , what is the probability  $M_n$  that  $\mathcal{G}_{n,p}$  is **not** a complete graph?



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For  $p = \frac{1}{2}$ , what is the probability  $M_n$  that  $\mathcal{G}_{n,p}$  is **not** a complete graph?

Any missing edge prevents  $\mathcal{G}_{n,p}$  from being complete, so

$$M_n = 1 - \left(\frac{1}{2}\right)^{\binom{n}{2}}.$$



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$M_n$	0%	50%	87.5%	98.4375%	$\geq 99.9\%$



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If  $P_n$  is the probability that property  $X$  holds for, say,  $\mathcal{G}_{n,p}$  and

$$\lim_{n \rightarrow \infty} P_n \rightarrow 1$$

then property  $X$  holds **asymptotically almost surely** or a.a.s.



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*What can we say asymptotically about  $\chi(\mathcal{G}_{n,p=d/n})$  or  $\chi(\mathcal{G}_{n,d})$ ?*





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- Erdős-Rényi (1960): Problem introduced
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- Kemkes-Pérez-Wormald (2009):  $\mathcal{G}_{n,d}$  concentrated on the same two values.



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- Coja-Oghlan et al. (2013): Use ideas from statistical physics to show both models concentrated on one value (for  $d$  large enough).



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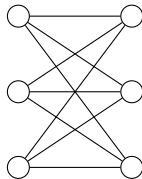
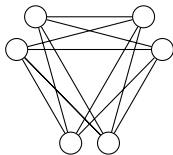
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## Key Idea (Overlap Matrices)

*In order to calculate second moments, need to know which types of graphs permit many colourings.*







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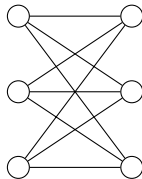
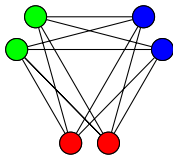
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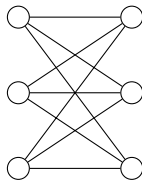
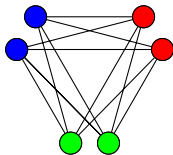
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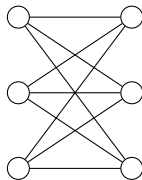
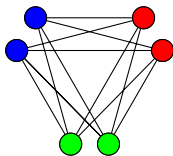
Motivation

Undirected  
Graphs

Directed  
Graphs

## Key Idea (Overlap Matrices)

*In order to calculate second moments, need to know which types of graphs permit many colourings.*



$$\begin{matrix} r & r & :0 \\ g & r & :0 \\ b & r & : \frac{1}{3} \end{matrix}$$

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# Undirected Graphs

A New  
Direction

JD Nir

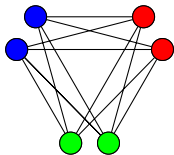
Motivation

Undirected  
Graphs

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Graphs

## Key Idea (Overlap Matrices)

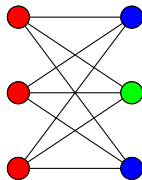
*In order to calculate second moments, need to know which types of graphs permit many colourings.*



$$\begin{matrix} \text{r r} & \text{g r} & \text{b r} \\ :0 & :0 & : \frac{1}{3} \end{matrix}$$

$$\begin{matrix} \text{r g} & \text{g g} & \text{b g} \\ : \frac{1}{3} & :0 & :0 \end{matrix}$$

$$\begin{matrix} \text{r b} & \text{g b} & \text{b b} \\ :0 & : \frac{1}{3} & :0 \end{matrix}$$



$$\begin{matrix} \text{r r} & \text{g r} & \text{b r} \\ : \frac{1}{6} & :0 & :0 \end{matrix}$$

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$$\begin{matrix} \text{r b} & \text{g b} & \text{b b} \\ : \frac{1}{3} & :0 & :0 \end{matrix}$$



# Undirected Graphs

A New Direction

JD Nir

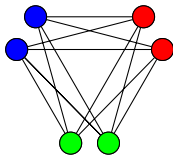
Motivation

Undirected Graphs

Directed Graphs

## Key Idea (Overlap Matrices)

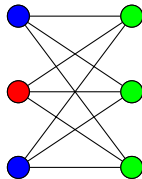
*In order to calculate second moments, need to know which types of graphs permit many colourings.*



$$\begin{matrix} \text{r r} & \text{g r} & \text{b r} \\ :0 & :0 & : \frac{1}{3} \end{matrix}$$

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# Undirected Graphs

A New  
Direction

JD Nir

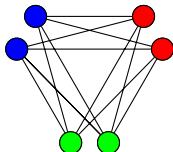
Motivation

Undirected  
Graphs

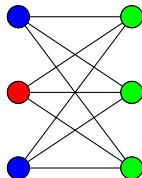
Directed  
Graphs

## Key Idea (Overlap Matrices)

*In order to calculate second moments, need to know which types of graphs permit many colourings.*



$$\begin{bmatrix} 0 & 0 & \frac{1}{3} \\ \frac{1}{3} & 0 & 0 \\ 0 & \frac{1}{3} & 0 \end{bmatrix}$$



$$\begin{bmatrix} \frac{1}{6} & 0 & 0 \\ 0 & \frac{1}{6} & \frac{1}{3} \\ \frac{1}{3} & 0 & 0 \end{bmatrix}$$



# Undirected Graphs

A New  
Direction

JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

## Key Idea (Overlap Matrices)

*In order to calculate second moments, need to know which types of graphs permit many colourings.*

Maximize  $f(\rho) = H(\rho) + E(\rho)$  where entropy and energy compete with  $\rho$  subject to constraints:



# Undirected Graphs

A New  
Direction

JD Nir

Motivation

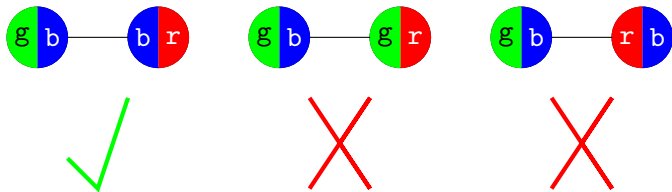
Undirected  
Graphs

Directed  
Graphs

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# Directed Graphs

## Oriented Colourings

A New  
Direction

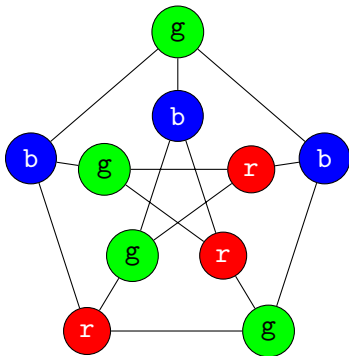
JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

### Graph Colouring





# Directed Graphs

## Oriented Colourings

A New  
Direction

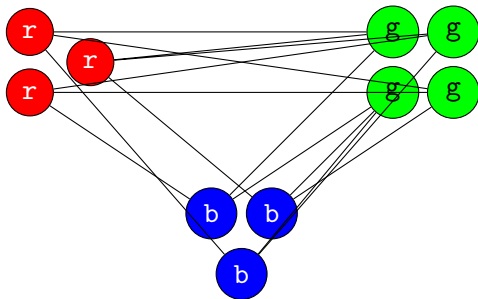
JD Nir

Motivation

Undirected  
Graphs

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Graphs

### Graph Colouring





# Directed Graphs

## Oriented Colourings

A New  
Direction

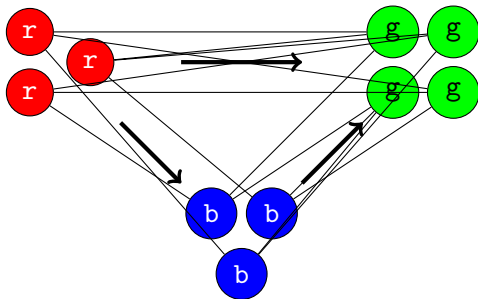
JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

### Oriented Graph Colouring





# Directed Graphs

## Oriented Colourings

A New  
Direction

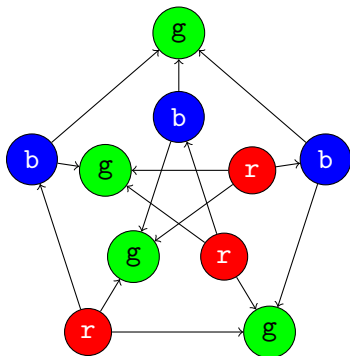
JD Nir

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### Oriented Graph Colouring





# Directed Graphs

## Oriented Colourings

A New  
Direction

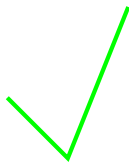
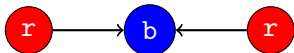
JD Nir

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Undirected  
Graphs

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Oriented Graph Colouring





# Directed Graphs

## Oriented Colourings

A New  
Direction

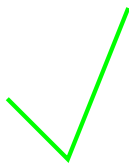
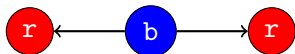
JD Nir

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Undirected  
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Oriented Graph Colouring





# Directed Graphs

## Oriented Colourings

A New  
Direction

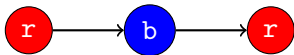
JD Nir

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Undirected  
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Directed  
Graphs

Oriented Graph Colouring





# Directed Graphs

## Oriented Colourings

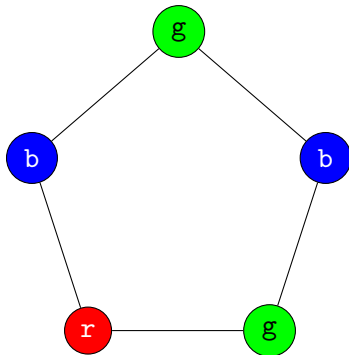
A New  
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Undirected  
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Directed  
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# Directed Graphs

## Oriented Colourings

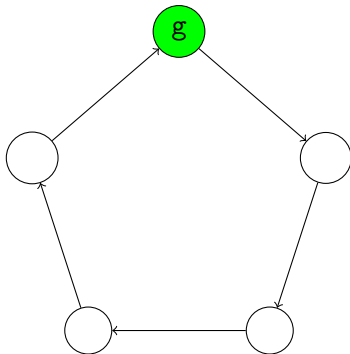
A New  
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Undirected  
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# Directed Graphs

## Oriented Colourings

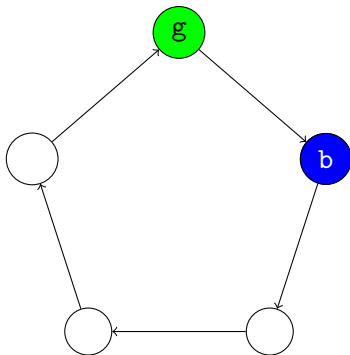
A New  
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Undirected  
Graphs

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Graphs





# Directed Graphs

## Oriented Colourings

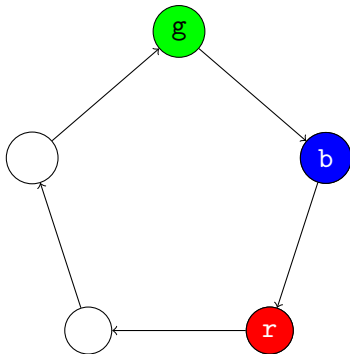
A New  
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# Directed Graphs

## Oriented Colourings

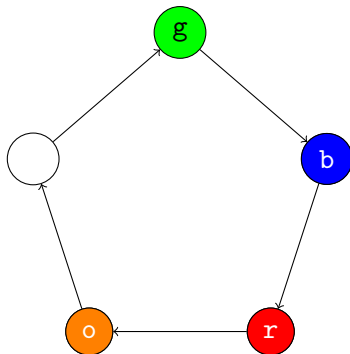
A New  
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# Directed Graphs

## Oriented Colourings

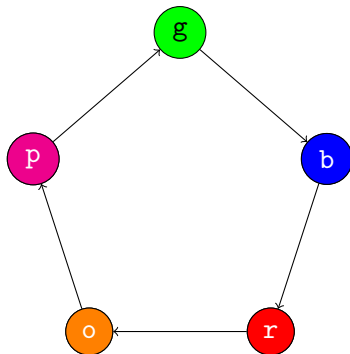
A New  
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# Directed Graphs

## Oriented Colourings

A New  
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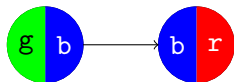
JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

Overlap matrices become more complicated:





# Directed Graphs

## Oriented Colourings

A New  
Direction

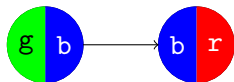
JD Nir

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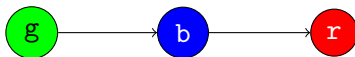
Undirected  
Graphs

Directed  
Graphs

Overlap matrices become more complicated:



requires





# Directed Graphs

## Doubly Regular Tournaments

A New  
Direction

JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

Question

*Which tournaments produce good product tournaments?*





# Directed Graphs

## Doubly Regular Tournaments

A New  
Direction

JD Nir

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Undirected  
Graphs

Directed  
Graphs

### Question

*Which tournaments produce good product tournaments?*

*Doubly regular* tournaments satisfy:

①  $d^+(v) = d^+(u)$  for every  $u, v \in V(\vec{G})$



# Directed Graphs

## Doubly Regular Tournaments

A New  
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JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

### Question

*Which tournaments produce good product tournaments?*

*Doubly regular* tournaments satisfy:

- 1  $d^+(v) = d^+(u)$  for every  $u, v \in V(\vec{G})$
- 2 Every pair of vertices  $u, v$  have the same number of common out-neighbours.



# Directed Graphs

## Doubly Regular Tournaments

A New  
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JD Nir

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Undirected  
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### Question

*Which tournaments produce good product tournaments?*

*Doubly regular* tournaments satisfy:

- 1  $d^+(v) = d^+(u)$  for every  $u, v \in V(\vec{G})$
- 2 Every pair of vertices  $u, v$  have the same number of common out-neighbours.

The product of a doubly-regular tournament with itself is strongly regular and has (unsigned) adjacency matrix

$$\frac{1}{2}(M \otimes M + (J - I) \otimes (J - I)).$$



# Directed Graphs

## Matrix Optimization

A New  
Direction

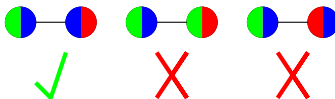
JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

More intricate constraints on overlap matrices:





# Directed Graphs

## Matrix Optimization

A New  
Direction

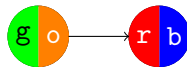
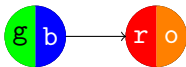
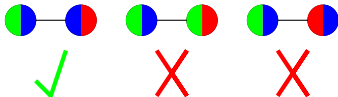
JD Nir

Motivation

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Directed  
Graphs

More intricate constraints on overlap matrices:





# Directed Graphs

## Results

A New  
Direction

JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

### Theorem (Gunderson-N., 2022+)

*The oriented chromatic numbers  $\chi_o(\vec{\mathcal{G}}_{n,p=d/n})$  and  $\chi_o(\vec{\mathcal{G}}_{n,d})$  are concentrated in the window*



# Directed Graphs

## Results

A New  
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Graphs

### Theorem (Gunderson-N., 2022+)

*The oriented chromatic numbers  $\chi_o(\vec{\mathcal{G}}_{n,p=d/n})$  and  $\chi_o(\vec{\mathcal{G}}_{n,d})$  are concentrated in the window*

$$(2^{d/2}, 6e^{d/2} + 6d + 17].$$



# Directed Graphs

## Results

A New  
Direction

JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

Maybe the exponential gap isn't our fault?





# Directed Graphs

## Results

A New  
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Maybe the exponential gap isn't our fault?

### Theorem

*For  $\vec{G} \sim \vec{G}_{n,2}$ , with high probability,  $\chi_o(\vec{G}) \in \{4, 5\}$ , but each occurs with positive probability.*



# Directed Graphs

## Results

A New  
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JD Nir

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Maybe the exponential gap isn't our fault?

### Theorem

*For  $\vec{G} \sim \vec{G}_{n,2}$ , with high probability,  $\chi_o(\vec{G}) \in \{4, 5\}$ , but each occurs with positive probability.*

Proof idea: with positive probability,  $\vec{G}$  has no oriented 5-cycle.



# Directed Graphs

Where to next?

A New  
Direction

JD Nir

Motivation

Undirected  
Graphs

Directed  
Graphs

Next steps:

- Adapt statistical physics models to directed case.



# Directed Graphs

Where to next?

A New  
Direction

JD Nir

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Undirected  
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Directed  
Graphs

Next steps:

- Adapt statistical physics models to directed case.  
Challenge: finding the right type of colouring.



# Directed Graphs

Where to next?

A New  
Direction

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Undirected  
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Graphs

Next steps:

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Challenge: finding the right type of colouring.
- Lower bound on concentration window length?



# Directed Graphs

Where to next?

A New  
Direction

JD Nir

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Undirected  
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Directed  
Graphs

Next steps:

- Adapt statistical physics models to directed case. Challenge: finding the right type of colouring.
- Lower bound on concentration window length?
- Focus on small cases, like  $d = 3$ . Great workshop problem!



# Thanks!

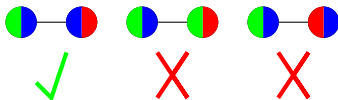
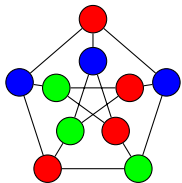
A New  
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JD Nir

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Undirected  
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Thank you!

