

# Load balancing and random graphs

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Includes joint work with Jane Gao

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**Theorem [Azar, Broder, Karlin and Upfal, '94]**

Throw the balls sequentially, each ball put into the least-full of  $h \geq 2$  randomly chosen bins.

Then finally, max is  $O(\ln \ln n / \ln h)$ .

# An application

Balance the load of machines, on-line.

'load' equals number of jobs assigned.

$h$  choices for each job.

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## Theorem [Berenbrink, Czumaj, Steger & Vöcking, '00]

With arbitrary  $m$ ,  
max load is a.a.s.  $m/n + O(\ln \ln n)$  (fixed  $h \geq 2$ ).



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Large array of  $n$  disks.

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files



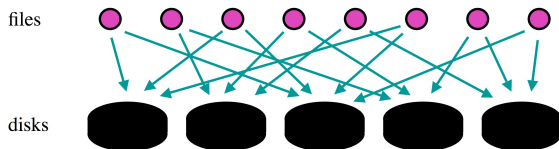
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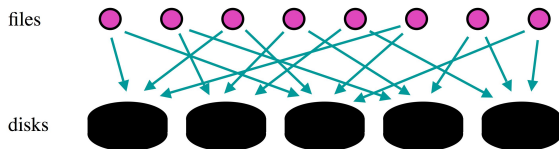


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Task: Balance the loads.

Backlog of processing → off-line load balancing

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Efficient ( $O(m^2)$ ) optimal algorithm exists — solving a max flow problem.

# Questions

(i) (Threshold) Fix  $k$ . What is the threshold  $m$  at which the maximum load will first exceed  $k$ .

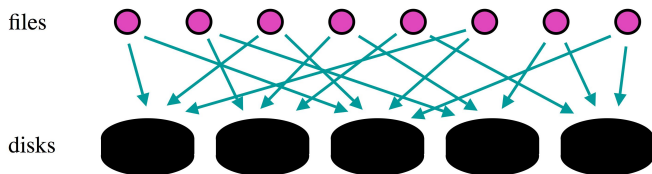
# Questions

- (i) (Threshold) Fix  $k$ . What is the threshold  $m$  at which the maximum load will first exceed  $k$ .
- (ii) (Approximate algorithms) What about faster algorithms that are 'nearly optimal', i.e. give the optimal load for most inputs with given parameters?

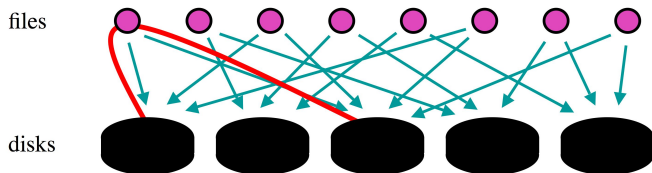


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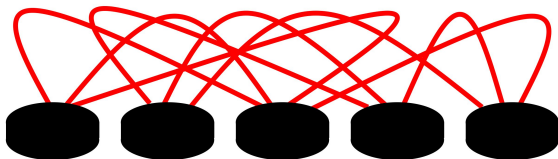
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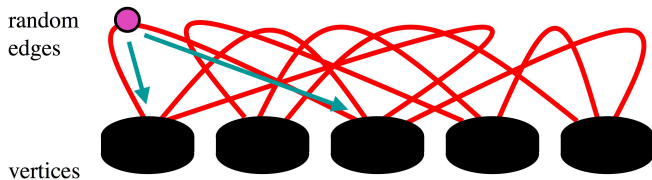
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files =  
edges

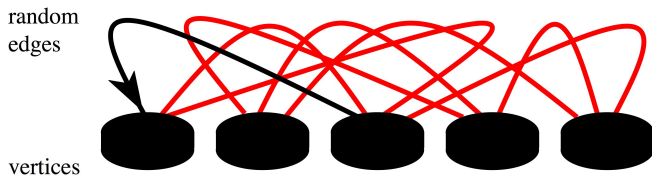
disks =  
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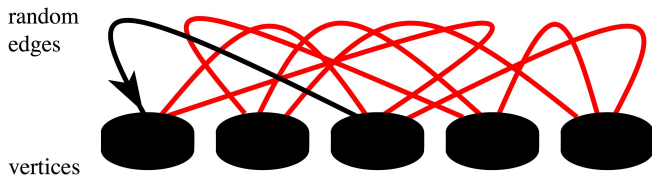
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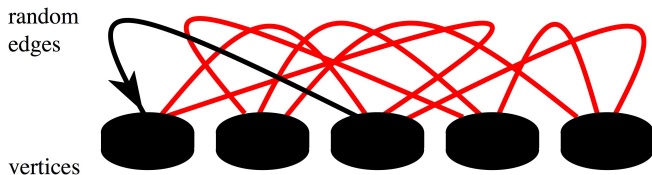


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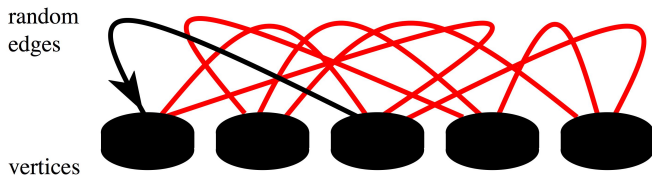


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**Assumption:** we consider  $G$  random in the space  $\mathcal{G}(n, m)$ .

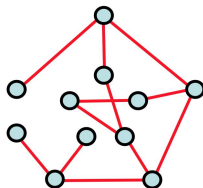
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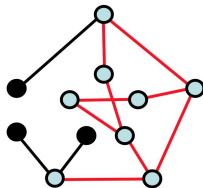
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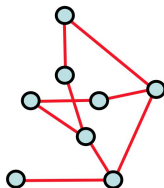
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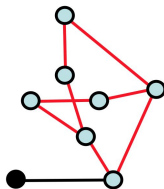
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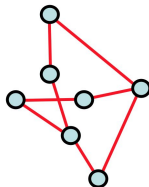
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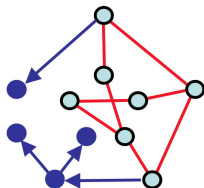
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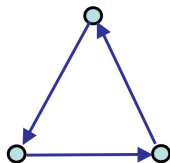
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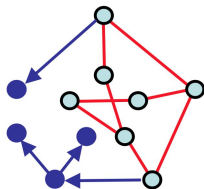
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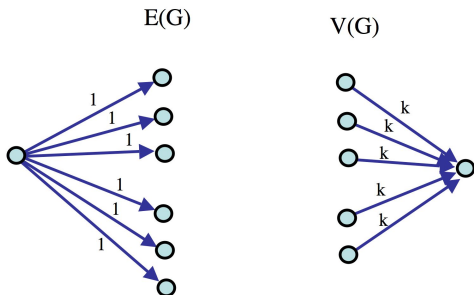
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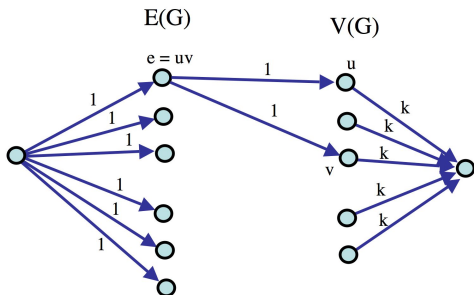




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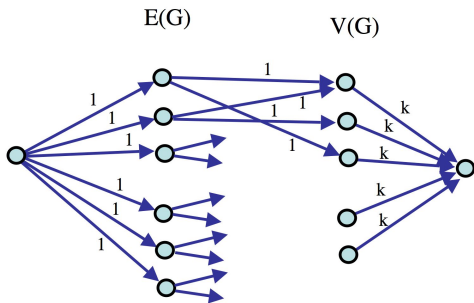
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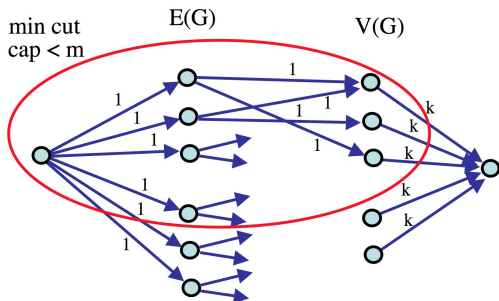
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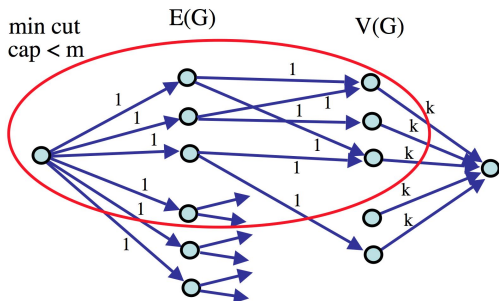
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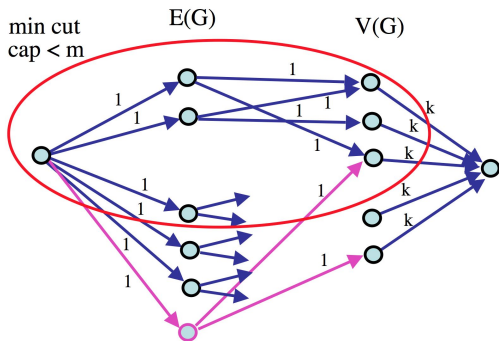
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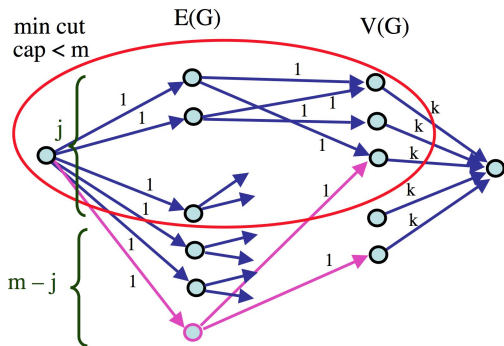
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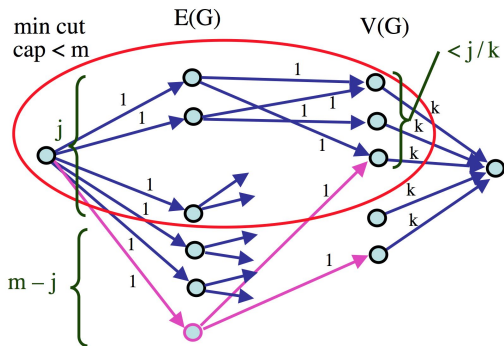
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$$c_k = k + d_k \sqrt{k} + O(1).$$

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This threshold for  $m$  is  $\mu_{k+1} n/2$  where  $\mu_{k+1}$  is constant.

$m < (\mu_{k+1} - \epsilon)n/2$  implies  $k$ -orientable (a.a.s.)

Proved algorithmically in both cases doing orientations edge by edge.

**Note:** we always orient the remaining incident edges to a vertex if doing so does not exceed load  $k$ .

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### load-degree algorithm

- \* orient edges one by one to vertices.
- \* a vertex with  $j$  edges oriented to it and  $i$  unoriented has 'load degree'  $j + i/2$ . Greedily choose a vertex of minimum load degree and orient an unoriented edge to it.



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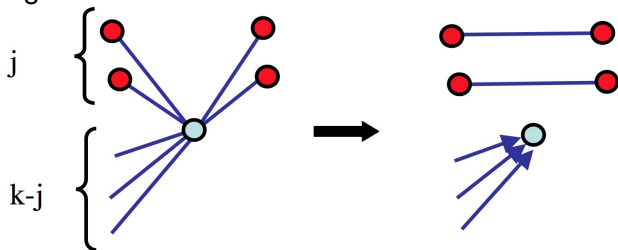
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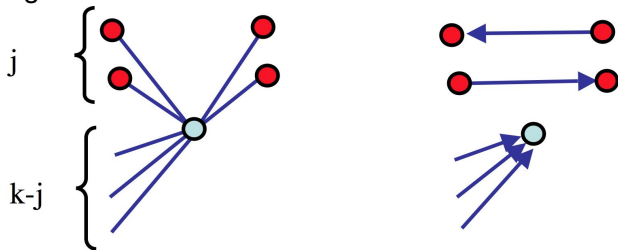
For  $c < \mu_{k+1} - \epsilon$ , the load-degree algorithm a.a.s. completes a  $k$ -orientation in a random graph  $\mathcal{G}(n, cn/2)$ .

FR's algorithm: choose a vertex of min degree,  $k + j$ , and do the following:

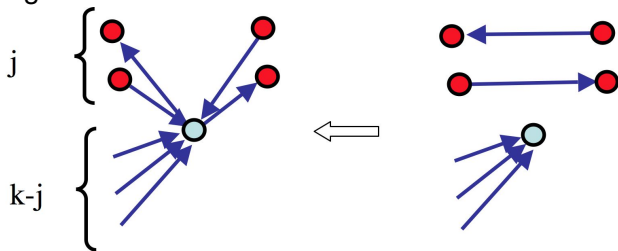
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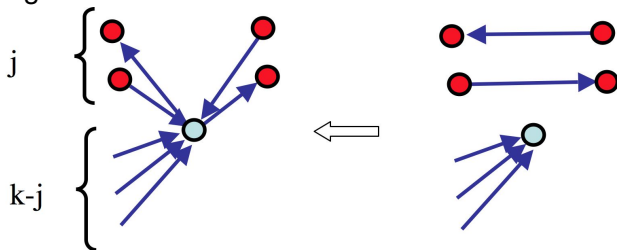
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The same flow proof gives:

Orientation exists iff no subhypergraph has average degree at least  $hk$ .

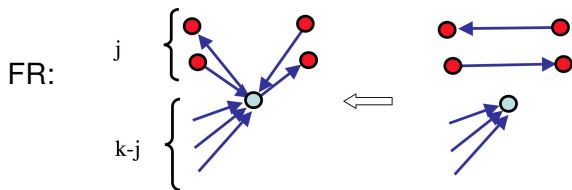
Clear first step: find the  $(k + 1)$ -core.

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**BOTH seem DIFFICULT!**



## Conjecture

*A.a.s. if  $\bar{d} < hk - \epsilon$ , where  $\bar{d}$  is the average degree in the  $(k+1)$ -core of  $G \in \mathcal{G}(n, m, h)$ , then  $G$  is  $k$ -orientable.*

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## Theorem [Gao and W, '09]

The conjecture is true for fixed  $k$  provided it is sufficiently large.

The threshold at which this occurs (core of  $h$ -uniform hypergraph gets average degree  $hk$ ) is known. (Cain and Wormald; Molloy, Cooper, Janson and Łuczak.)

## Problem version 3. Even more to choose from

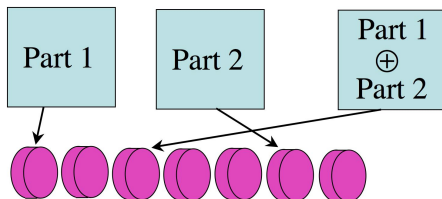
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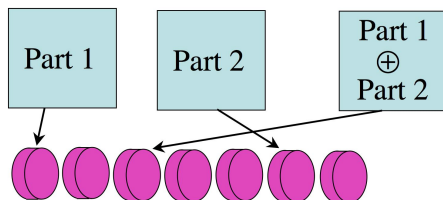
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... and what if two disks fail? More general error correcting strategies.

Each file will need to choose  $w$  out of  $h$  disks, for some  $1 \leq w < h$ .



# More general orientations

A hyperedge is  $w$ -oriented if exactly  $w$  distinct vertices in it are marked with positive signs with respect to the hyperedge.

The  $\text{indegree}$  of a vertex is the number of positive signs it receives.

A  $(w, k)$ -orientation of an  $h$ -hypergraph is a  $w$ -orientation all hyperedges such that each vertex has indegree at most  $k$ .

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The disks can be allocated in the  $w$ -out-of- $h$  setting iff the corresponding hypergraph is  $(w, k)$ -orientable.

# More general cores

**light** vertex: one with degree at most  $k$ .

**free step (I)** : For any light vertex  $v$ , give  $v$  positive sign with respect to all incident hyperedges, delete  $v$  and remove  $v$  from all incident hyperedges *but don't delete those hyperedges* .

**free step (II)** : remove any hyperedge if its cardinality falls to  $h - w$ . (Must then have  $w$  vertices with positive signs.)

# More general cores

**light** vertex: one with degree at most  $k$ .

**free step (I)** : For any light vertex  $v$ , give  $v$  positive sign with respect to all incident hyperedges, delete  $v$  and remove  $v$  from all incident hyperedges *but don't delete those hyperedges* .

**free step (II)** : remove any hyperedge if its cardinality falls to  $h - w$ . (Must then have  $w$  vertices with positive signs.)

**$(w, k)$ -core** : the result of repeating free steps until no more can be taken. Has mixed edge sizes.

## Conjecture

*The event*

$$\sum_{j=0}^{w-1} (w-j)m_{h-j} < k\bar{n},$$

*where*

$m_{h-j}$  is the number of edges of cardinality  $h-j$  in the  $(w, k+1)$ -core of  $G \in \mathcal{G}(n, m, h)$ ,

$\bar{n}$  is its number of vertices,

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has the same sharp threshold as  $G$  being  $(w, k)$ -orientable. In particular  $(w, k)$ -orientability has a sharp threshold.

## Theorem [Gao and W, '09]

The conjecture is true for fixed  $k$  sufficiently large.

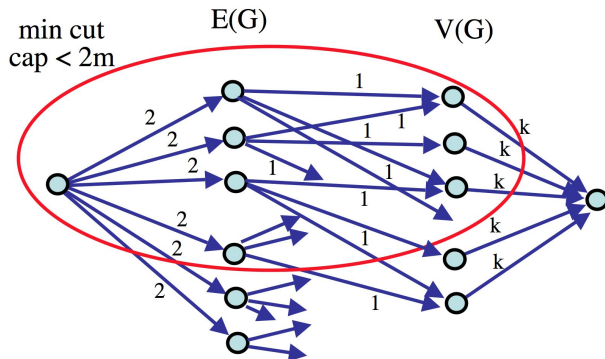
We determine the threshold in terms of the solution of a system of differential equations.

# Proof approach

Look again at the flow formulation. E.g.:  $h = 3$ ,  $w = 2$ .

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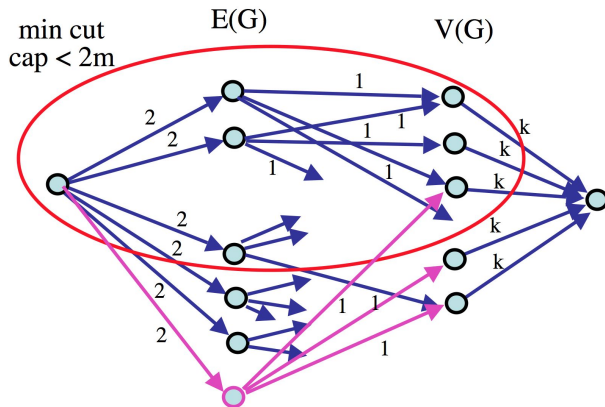
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From the flow:

### Lemma

An  $h$ -hypergraph  $G$  can be  $(w, k)$ -oriented iff there is no set  $S \subseteq V(G)$  with

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Major result:

### Sub-theorem

Let  $\mathcal{C}$  denote the  $(w, k + 1)$ -core of  $G \in \mathcal{G}(n, m, h)$ . A.a.s. if

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for  $S = \mathcal{C}$  then it holds for all  $S \subseteq V(G)$ .

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$$\frac{1}{|\mathcal{C}|} \sum_{j=0}^{w-1} (w-j)m_{h-j} < k - \epsilon$$

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This gives the theorem.

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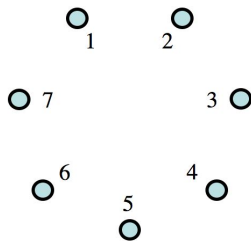
To study the threshold we also need properties of the random  $(w, k+1)$ -core.

# Analysing the random $(w, k + 1)$ -core

A random pseudograph model of Bollobás and Frieze ('85), Chvátal ('91), etc:

Given positive integers  $n$  and  $m$ , define the probability space of functions  $f : [2m] \rightarrow [n]$ , all functions equiprobable.

Pseudograph  $\mathcal{P}(n, m)$ : edges are  $\{f(2i - 1), f(2i)\}$  ( $1 \leq i \leq m$ ).

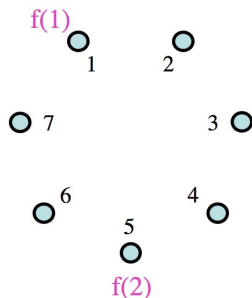


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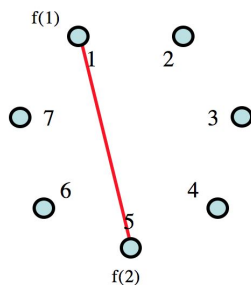


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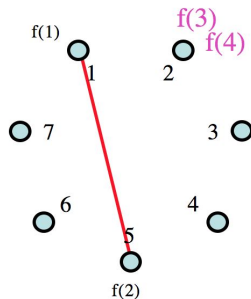


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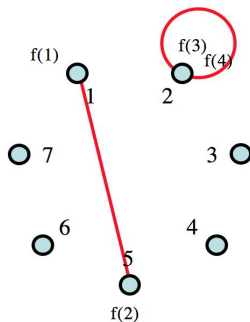


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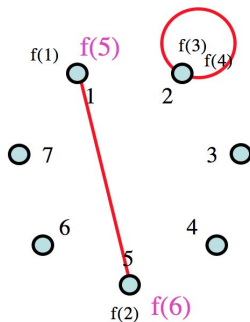


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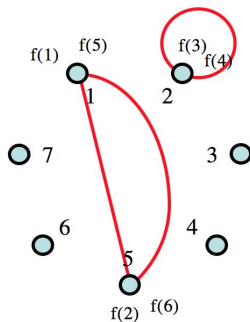


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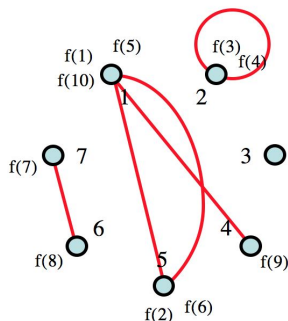


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- ▶ **Multinomial property** of degree sequence  $(X_1, \dots, X_n)$ : for nonnegative integer vector  $(d_1, \dots, d_n)$  with  $\sum_i d_i = 2m$ ,

$$\mathbf{P}(X_i = d_i \text{ for } 1 \leq i \leq n) = \frac{(2m)!}{n^{2m} \prod_{i=1}^n d_i!}.$$

# Our adaptation

For a random pseudo-hypergraph  $\mathcal{P}_n(h_1, \dots, h_m)$  with

- ▶  $n$  vertices
- ▶  $m$  edges of cardinalities  $h_1, \dots, h_m$

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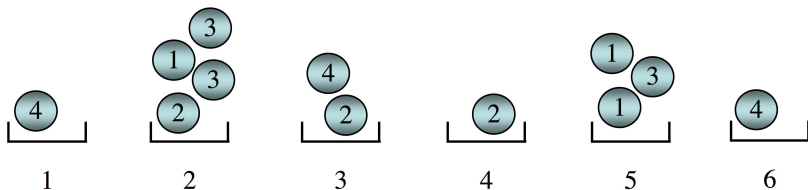
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# Finding the $(w, k + 1)$ -core of the pseudo-hypergraph

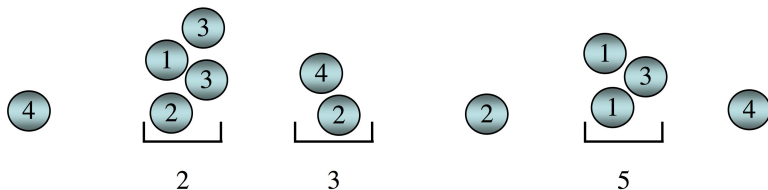
$h = 3$     $w = 2$     $k = 1$

$\{\text{balls}\} = \{1, 1, 1, 2, 2, 2, 3, 3, 3, 4, 4, 4\}$



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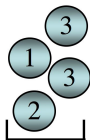


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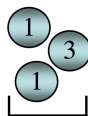
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2



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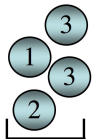
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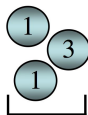
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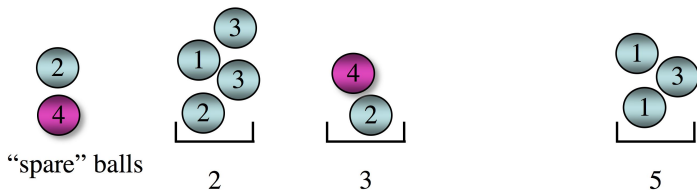


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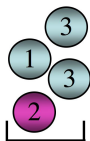


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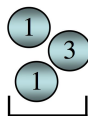
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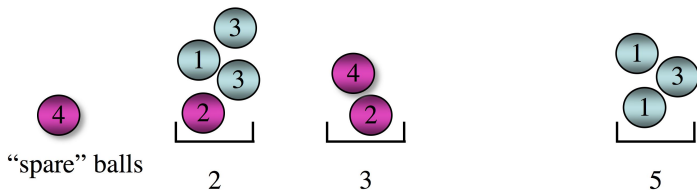
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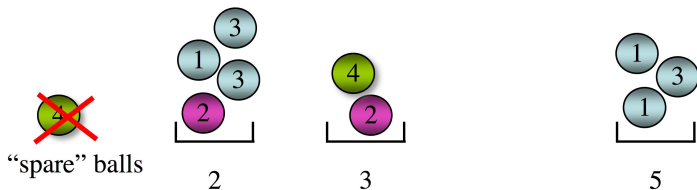
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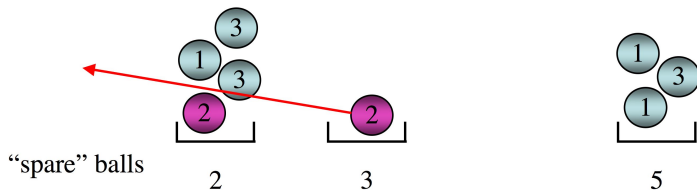
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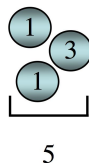
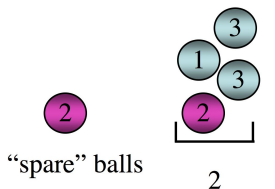
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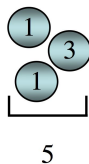
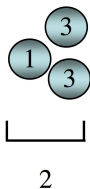
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The number of balls in the bins therefore has a ‘truncated’ multinomial distribution which is close to a set of copies of truncated Poisson:

$$\mathbf{P}(X = j) = \frac{e^{-\lambda} \lambda^j / j!}{\sum_{i \geq k+1} e^{-\lambda} \lambda^i / i!}.$$

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Even without d.e.'s we can prove (after considerable work, and for large  $k$ )

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- ▶ Conjecture the same results for all  $k \geq 1$ .
- ▶ No fast algorithm for loadbalancing yet. Conjecture that the obvious generalisation of the load-degree algorithm does the job for all  $(w, k)$ .