

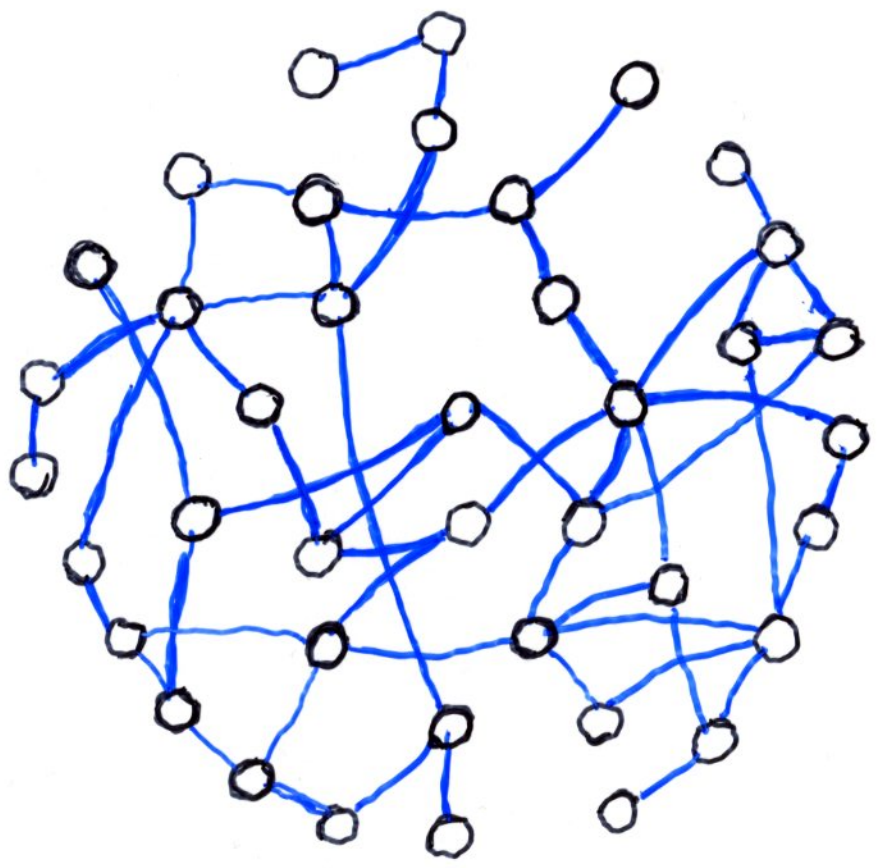
ON THE APPROXIMABILITY  
OF INDEPENDENT SET PROBLEM  
IN POWER LAW GRAPHS

MATHIAS HAUPTMANN

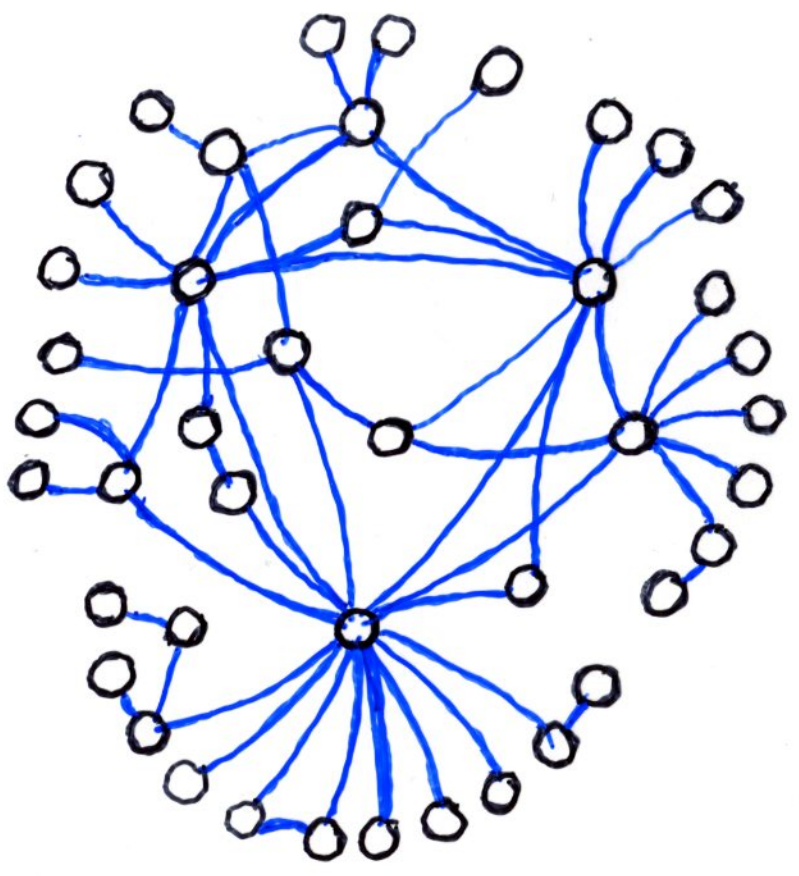
JOINT WORK WITH MAREK KARPINSKI

# POWER LAW GRAPHS

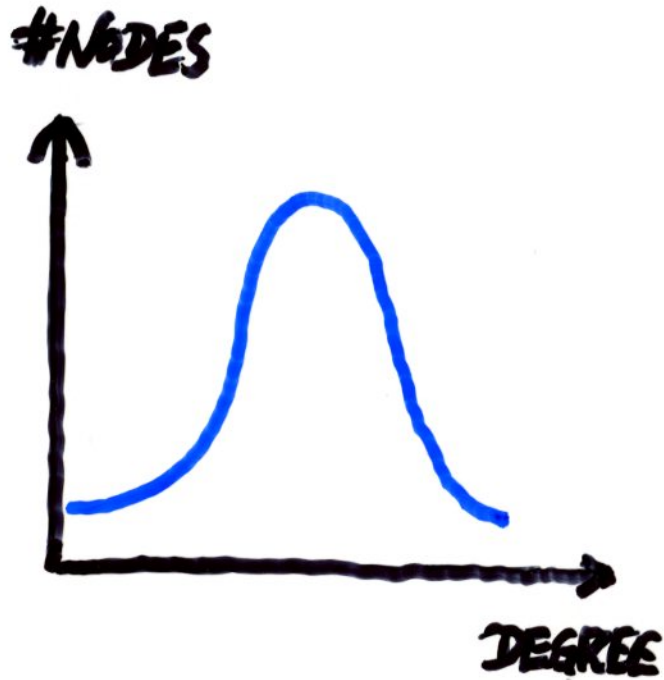
ERDŐS - RÉNYI



POWER LAW GRAPH (PLG)



# POWER LAW GRAPHS

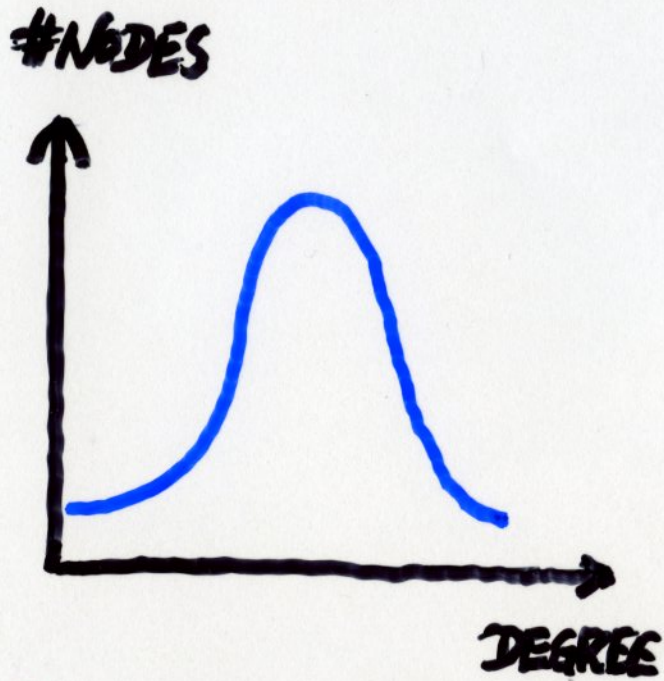


ERDŐS - RÉNYI

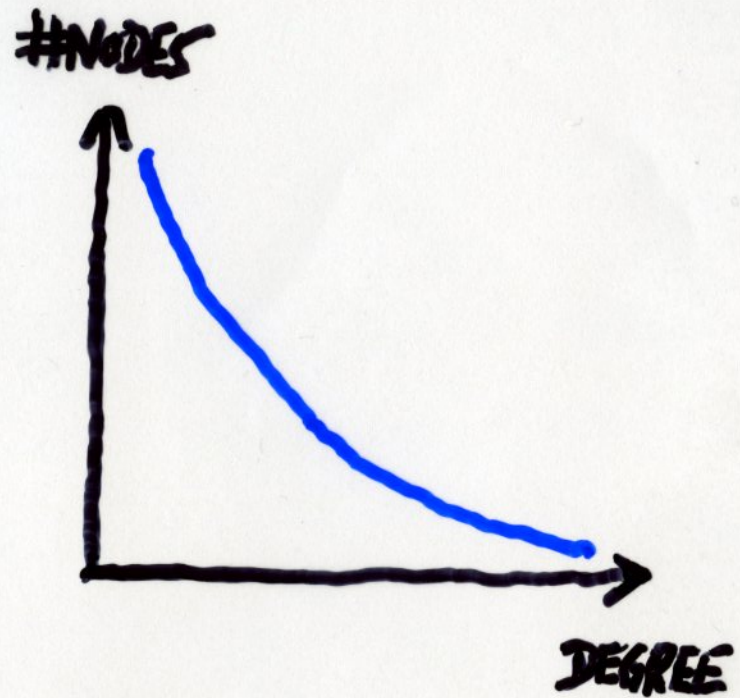


POWER LAW  
(SCALE-FREE)

# POWER LAW GRAPHS



ERDÖS - RÉNYI

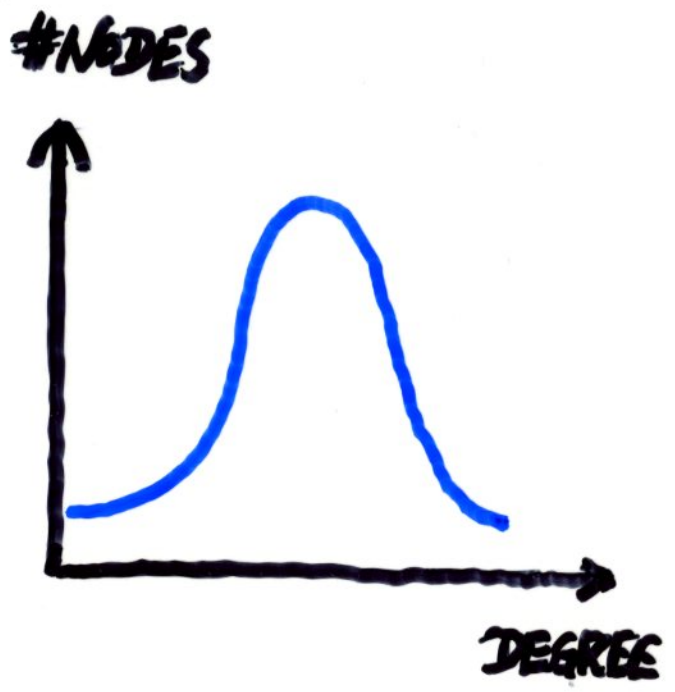


POWER LAW

(SCALE-FREE)

$$P(k) \sim k^{-\beta}$$

# POWER LAW GRAPHS



ERDÖS - RÉNYI



POWER LAW  
(SCALE-FREE)

$\beta =$  POWER LAW EXONENT  $\rightarrow$

$$P(k) \sim k^{-\beta}$$

# OUTLINE

- POWER LAW GRAPHS (PLG)
- MAXIMUM INDEPENDENT SET (MIS)
- NEW APPROXIMATION LOWER BOUNDS FOR MIS ON POWER LAW GRAPHS
  - ▷ AUXILIARY GRAPH CONSTRUCTION
  - ▷ CASE  $\beta < 1$
  - ▷ CASE  $\beta = 1$
- OPEN PROBLEMS

# EXAMPLES OF SCALE-FREE NETWORKS

- COMPUTER NETWORKS  
EG. WEBGRAPH OF THE WWW
- SOCIAL NETWORKS, COLLABORATION NETWORKS
- FINANCIAL NETWORKS  
EG. INTERBANK PAYMENT NETWORKS
- BIOLOGICAL NETWORKS
  - ▶ PROTEIN INTERACTION NETWORKS

# $(\alpha, \beta)$ -POWER LAW GRAPHS

MAX DEGREE

$$\Delta = \lfloor e^{\alpha/\beta} \rfloor$$

# NODES OF  
DEGREE  $i$ :

$$y_i = \lfloor \frac{e^\alpha}{i^\beta} \rfloor$$



# $(\alpha, \beta)$ -POWER LAW GRAPHS

MAX DEGREE  $\Delta = \lfloor e^{\alpha/\beta} \rfloor$

# NODES OF DEGREE  $i$ :  $y_i = \lfloor \frac{e^\alpha}{i^\beta} \rfloor$

# NODES	FOR
$n \approx \left\{ \begin{array}{l} \frac{e^{\alpha/\beta}}{1-\beta} \\ \alpha \cdot e^\alpha \\ \zeta(\beta) \cdot e^\alpha \end{array} \right.$	$0 < \beta < 1$
	$\beta = 1$
	$\beta > 1$

# $(\alpha, \beta)$ -POWER LAW GRAPHS

MAX DEGREE  $\Delta = \lfloor e^{\alpha/\beta} \rfloor$

# NODES OF DEGREE  $i$ :  $y_i = \lfloor \frac{e^\alpha}{i^\beta} \rfloor$

# NODES	FOR	# EDGES	FOR
$n \approx \frac{e^{\alpha/\beta}}{1-\beta}$	$0 < \beta < 1$	$\frac{1}{2} \frac{e^{2\alpha/\beta}}{2-\beta}$	$0 < \beta < 2$
$\alpha \cdot e^\alpha$	$\beta = 1$	$\frac{1}{4} \alpha e^\alpha$	$\beta = 2$
$\zeta(\beta) \cdot e^\alpha$	$\beta > 1$	$\frac{1}{2} \zeta(\beta-1) e^\alpha$	$\beta > 2$

# RANDOM PLGs

[ACLO1]

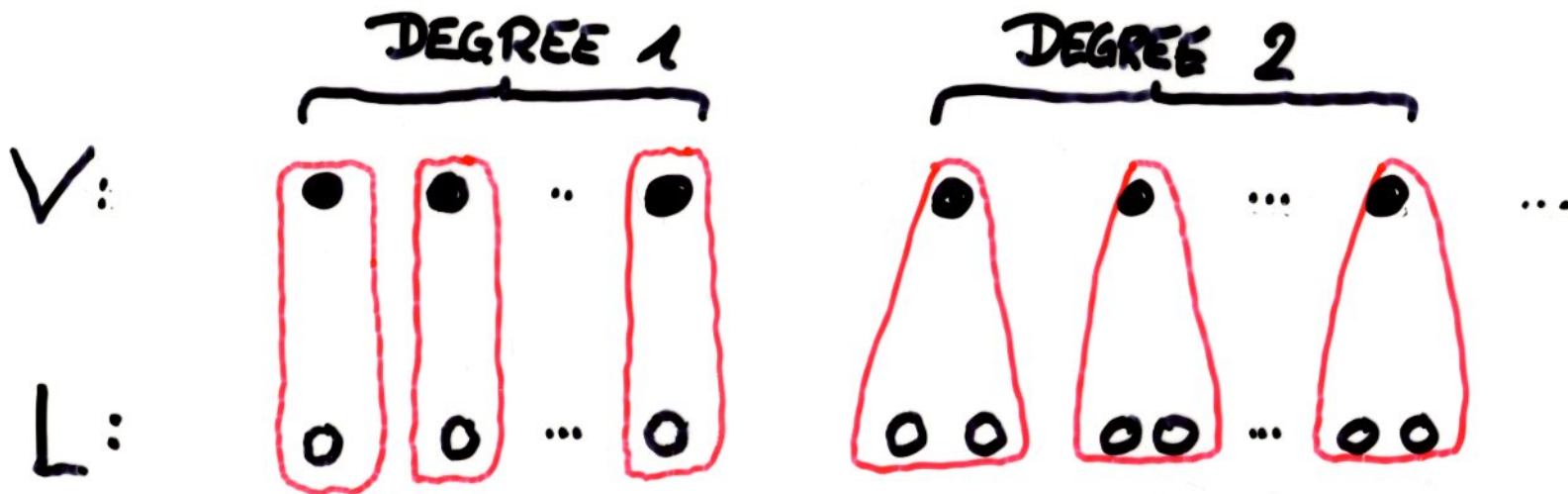
1. GENERATE SET  $L$  OF  $d(v)$  COPIES FOR EACH  $v \in V$
2.  $M :=$  RANDOM PERFECT MATCHING ON  $L$
3.  $M$  INDUCES (MULTI-)GRAPH  $G$



# RANDOM PLGs

[ACLO1]

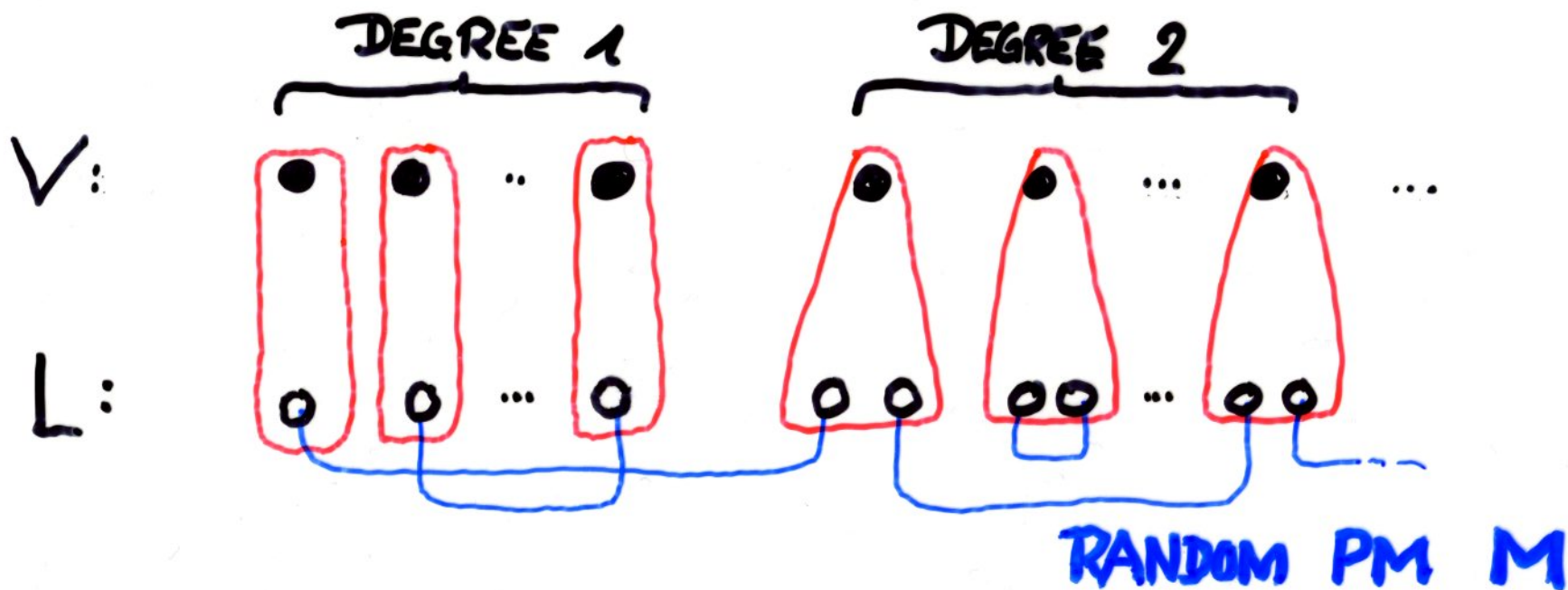
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# RANDOM PLGs

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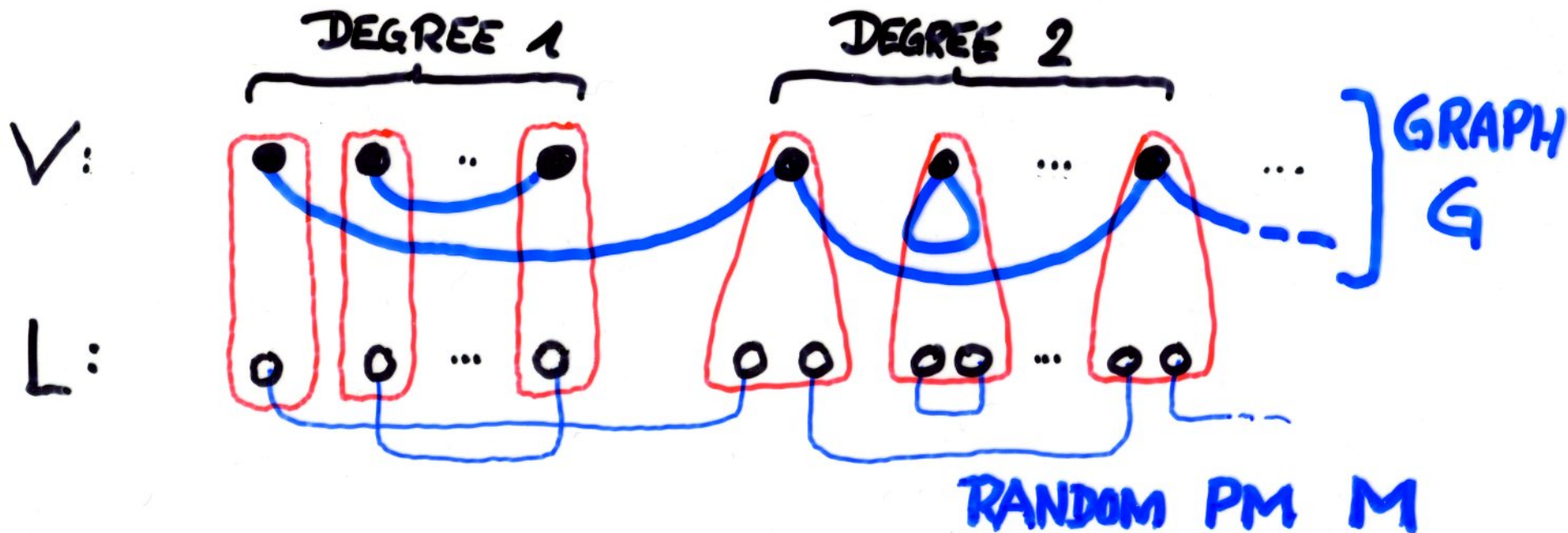
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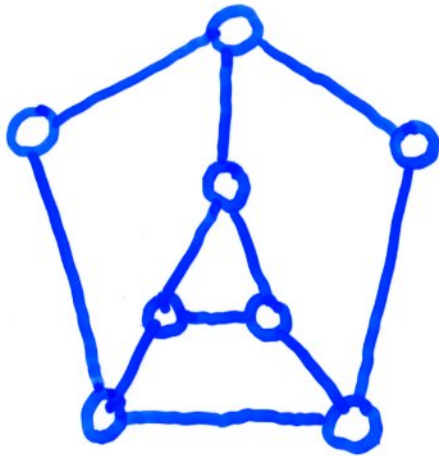
# RANDOM PLGs

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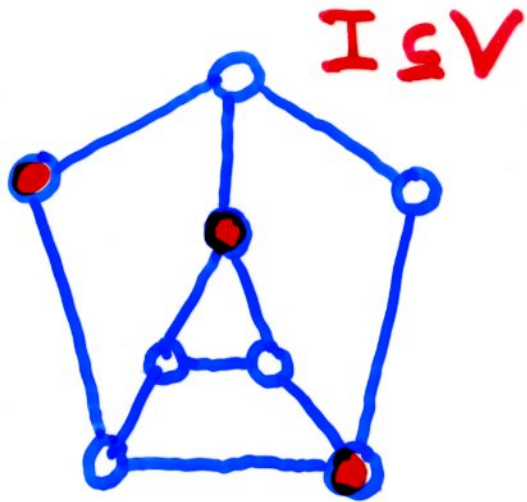


# MAXIMUM INDEPENDENT SET (MIS)



GRAPH G

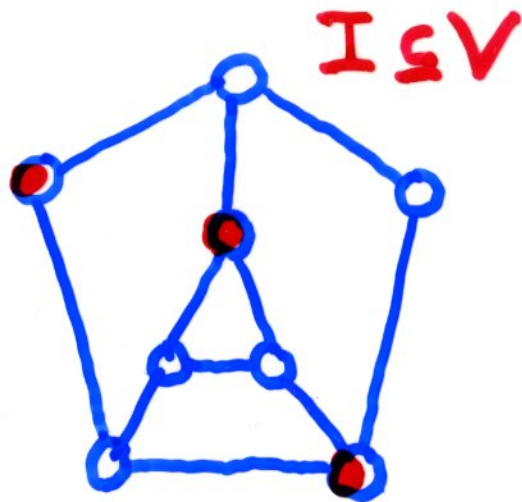
# MAXIMUM INDEPENDENT SET (MIS)



GRAPH G



# MAXIMUM INDEPENDENT SET (MIS)



## STATUS OF MIS

- ▶ MIS EQUIVALENT TO MAX-CLIQUE
- ▶ NP-HARD TO APPROX. WITHIN  $n^{1-\epsilon}$   
(FOR EVERY  $\epsilon > 0$ ) [Z07]
- ▶ NOT APPROXIMABLE WITHIN  $n / 2(\log n)^{1-\delta}$  UNLESS  $\tilde{NP} = \tilde{P}$

# MIS IN COMPLEX NETWORKS

- SCHEDULING IN WIRELESS SENSOR NETWORKS  
EFFICIENTLY EMPTYING WSN
- SECONDARY STRUCTURE MOTIFS IN PROTEINS
- COMMUNITY STRUCTURE IN COMPLEX NETWORKS
- EPIDEMICS

# MIS IN POWER LAW GRAPHS

## PREVIOUS RESULTS

- NP-HARD FOR ALL  $\beta > 0$  (Ferrante et al. '08)
- APX-HARD FOR ALL  $\beta \geq 1$  (Sten et al. '12.)  
EXPLICIT APPROX. LOWER BOUNDS

# MIS IN POWER LAW GRAPHS

## PREVIOUS RESULTS

- NP-HARD FOR ALL  $\beta > 0$  (Ferrante et al. '08)
  - APX-HARD FOR ALL  $\beta > 1$  (Sten et al. '12)
- EXPLICIT APPROX. LOWER BOUNDS

## OUR RESULTS

- FOR  $\beta < 1$ , NP-HARD TO APPROX.  
WITHIN  $n^{1-\beta-\epsilon}$  FOR ALL  $\epsilon > 0$
- FOR  $\beta = 1$ , LOWER BOUND  $(\log n)^\epsilon$   
FOR SOME  $\epsilon \in (0, 1)$

# THE CASE $\beta < 1$

## OUTLINE

# THE CASE $\beta < 1$

## OUTLINE

$n^{1-\epsilon}$  LOWER BOUND  
FOR MIS (ZUCKERMAN'07)

# THE CASE $\beta < 1$

## OUTLINE

CLASS OF GRAPHS

$$\mathcal{G} = \mathcal{G}_n^\varepsilon \cup \mathcal{G}_n^{1-\varepsilon}$$

$n^{1-\varepsilon}$  LOWER BOUND

FOR MIS (ZUCKERMAN'07)

# THE CASE $\beta < 1$

## OUTLINE

CLASS OF GRAPHS

$$\mathcal{G} = \mathcal{G}_{n^\varepsilon} \cup \mathcal{G}_{n^{1-\varepsilon}} \ni G \mapsto G'$$

STEP 1

HAS  
PERF. MATCHING

$n^{1-\varepsilon}$  LOWER BOUND

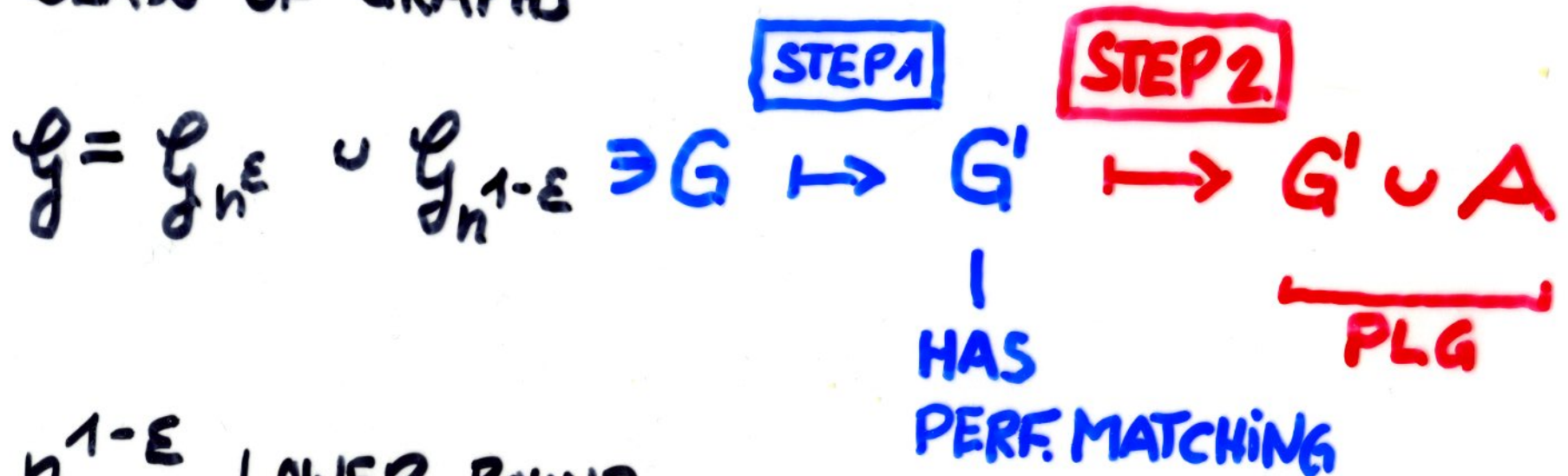
FOR MIS (ZUCKERMAN'07)



# THE CASE $\beta < 1$

## OUTLINE

CLASS OF GRAPHS

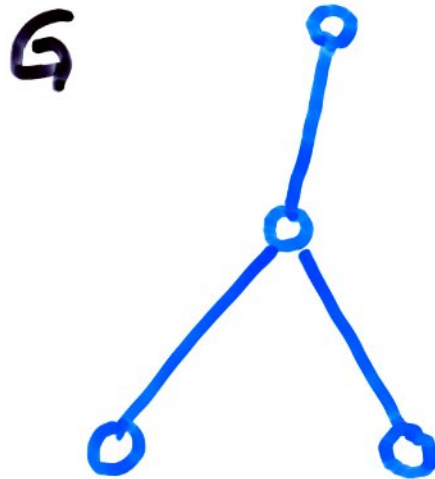


$n^{1-\varepsilon}$  LOWER BOUND

FOR MIS (ZUCKERMAN'07)

# EMBEDDING OF GRAPHS INTO POWER LAW GRAPHS

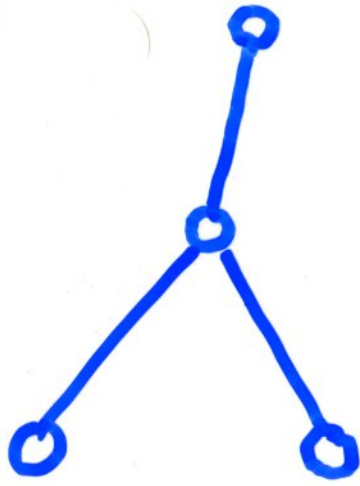
STEP 1  $G \mapsto G'$  SUCH THAT  $G'$  HAS  
PERFECT MATCHING



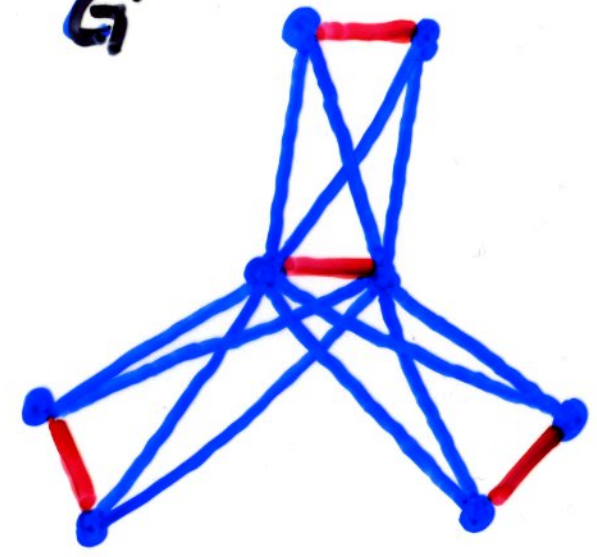
# EMBEDDING OF GRAPHS INTO POWER LAW GRAPHS

STEP 1  $G \mapsto G'$  SUCH THAT  $G'$  HAS  
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$G$



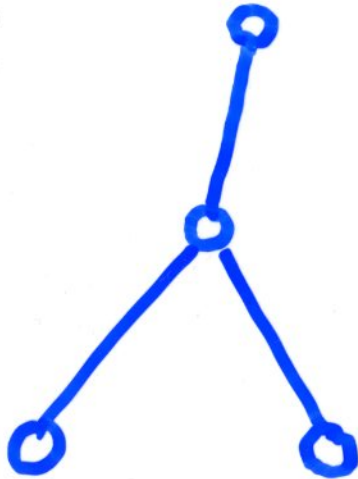
$G'$



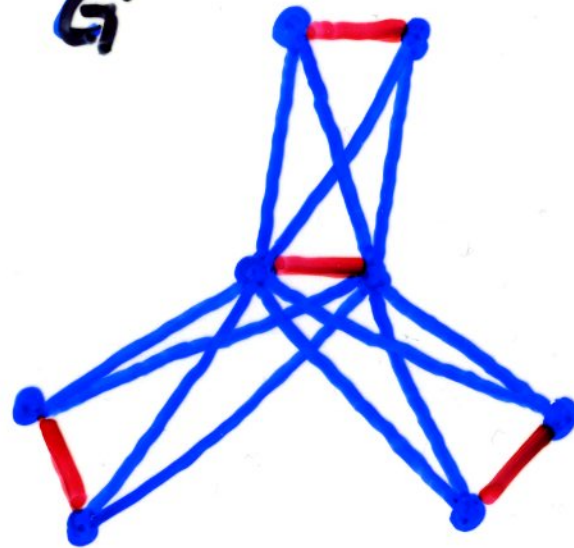
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STEP 1  $G \mapsto G'$  SUCH THAT  $G'$  HAS  
PERFECT MATCHING

$G$



$G'$



PROPERTY:  $IS(G') = IS(G)$

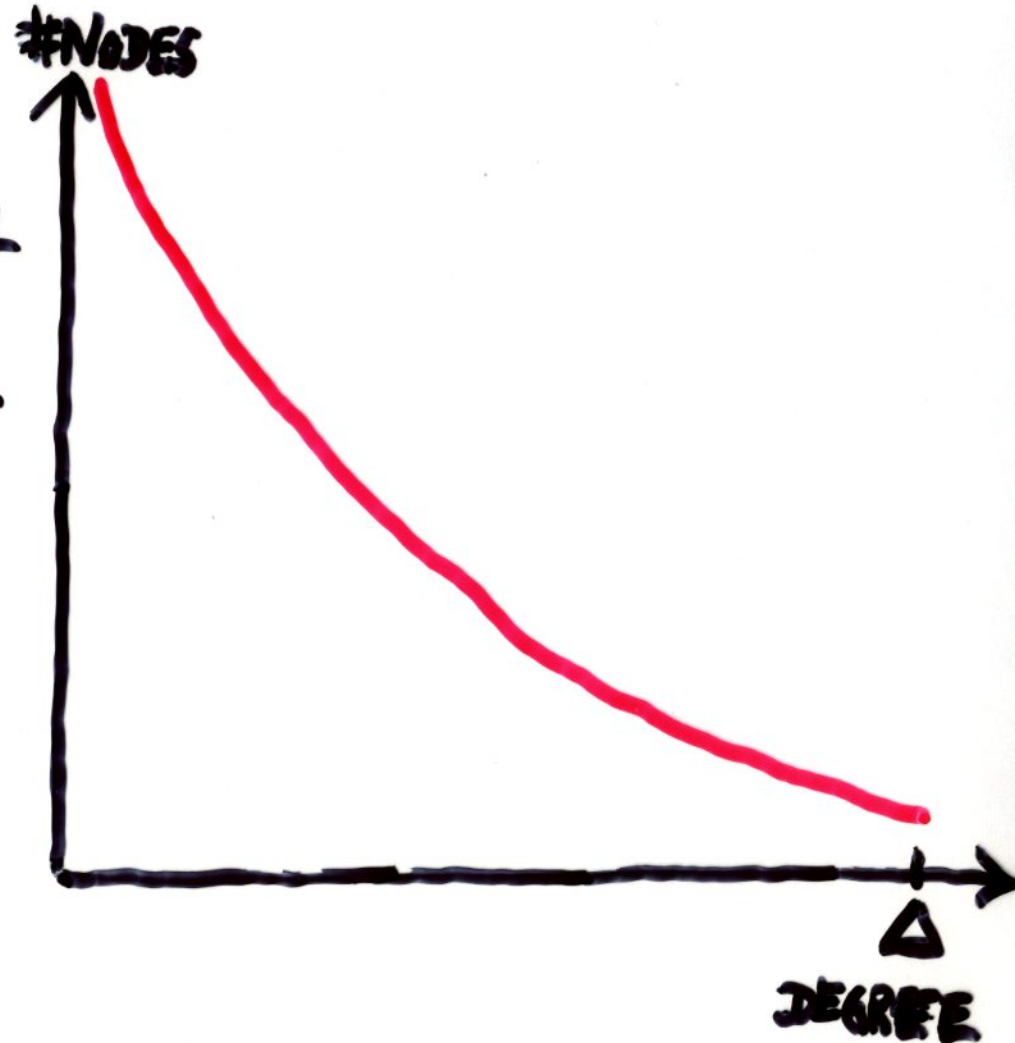
# EMBEDDING OF $G'$ INTO PLG

- CASE  $\beta < 1$  -

▷ USE PERFECT MATCHING  
TO FIT  $G'$  INTO INTERVAL  
 $[x\Delta, \Delta]$  OF PL DISTRIB.

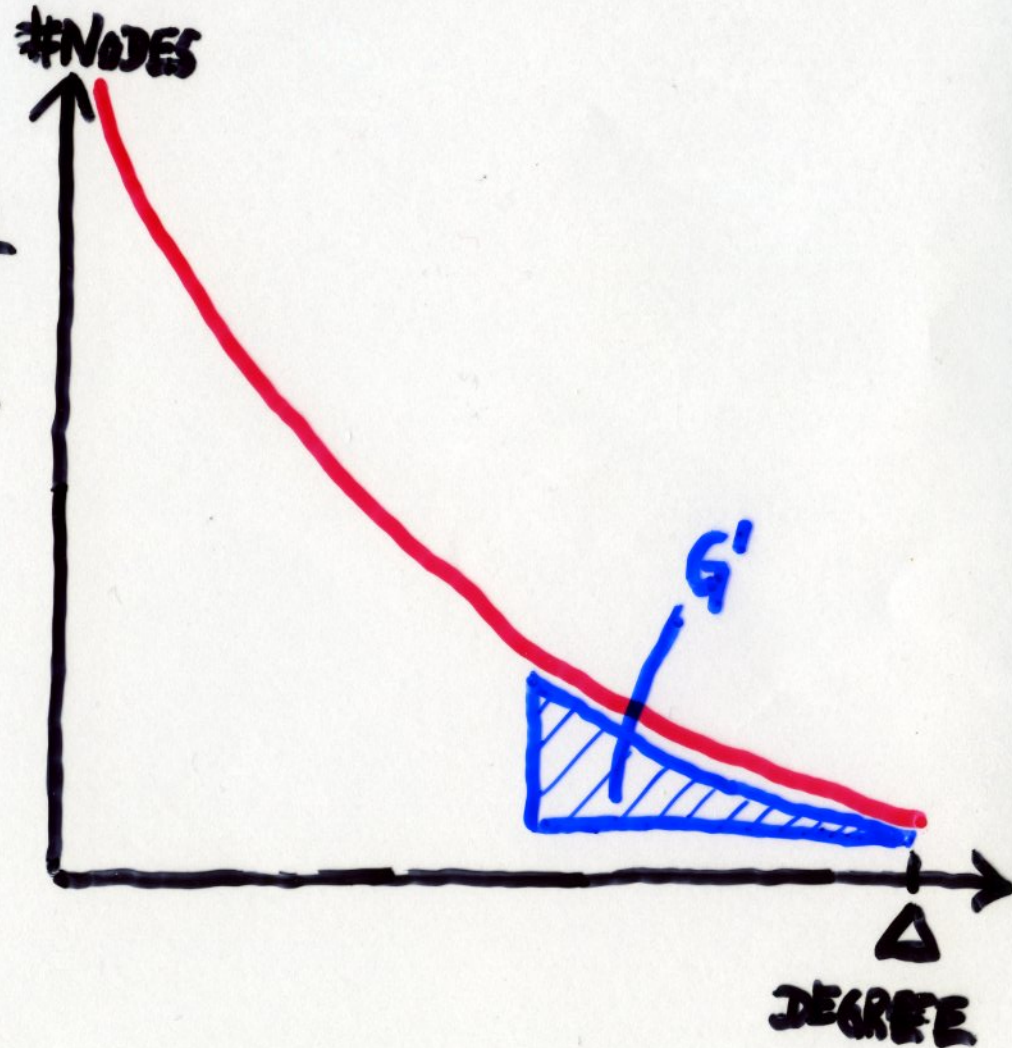
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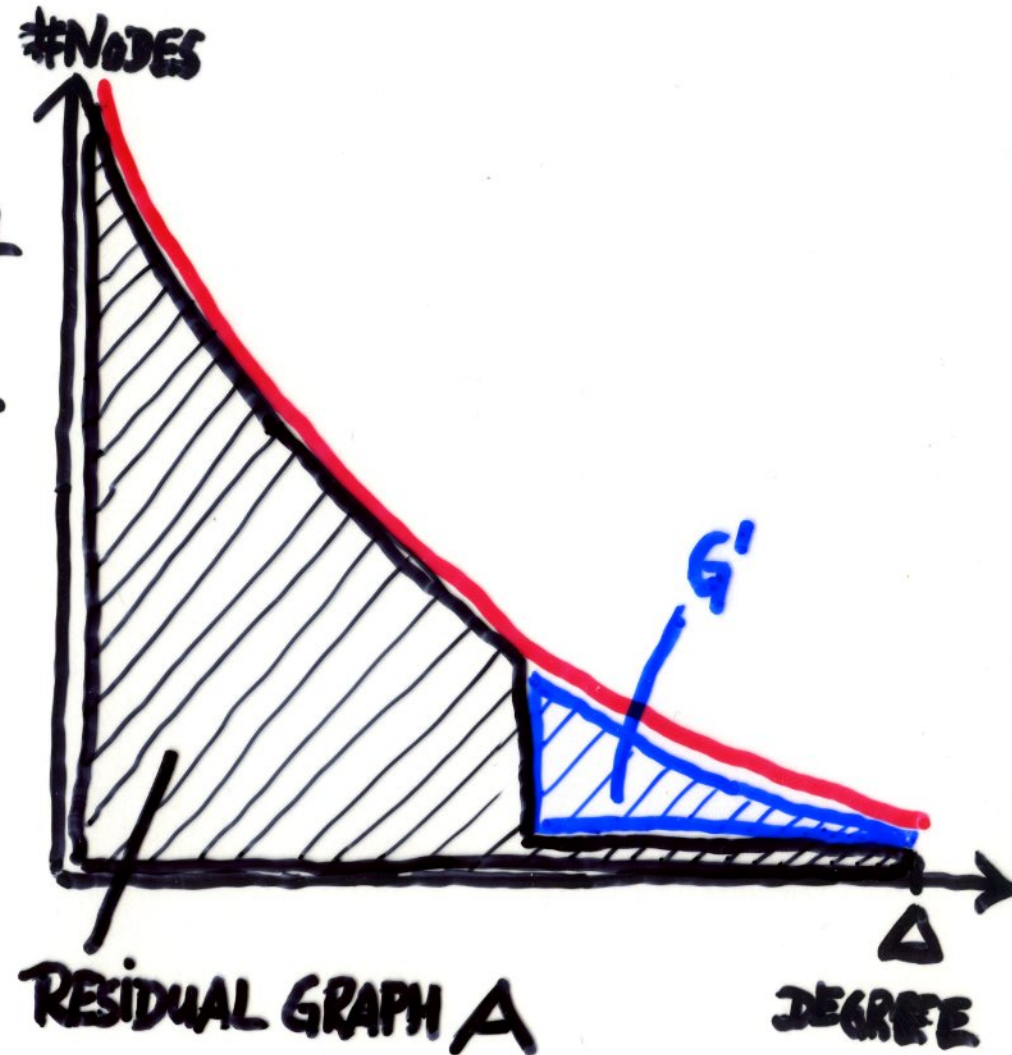
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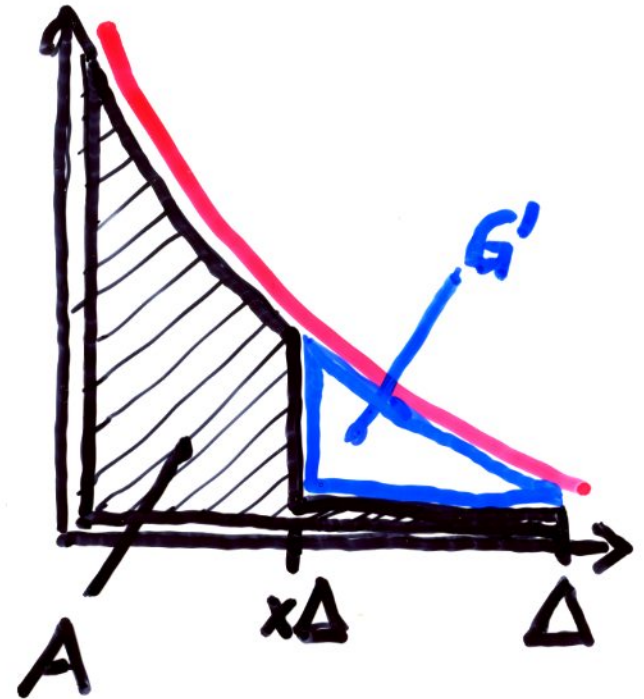
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# CONSTRUCTION OF A



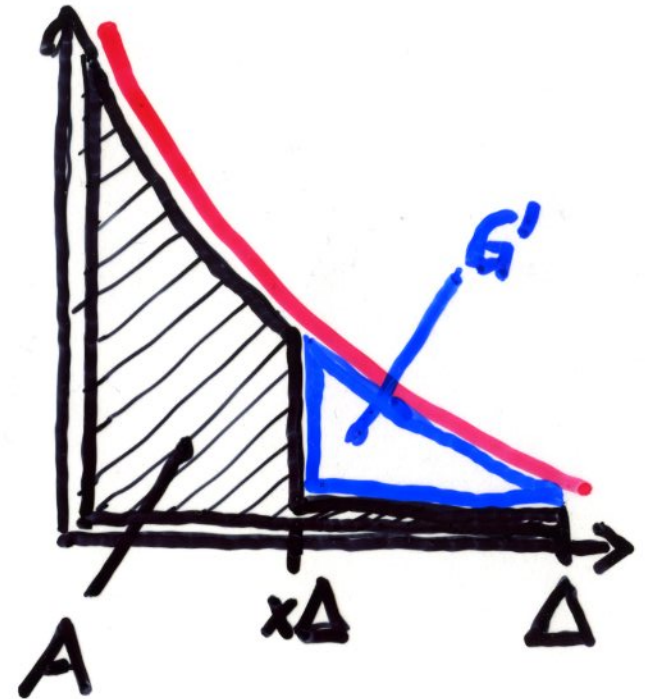
# CONSTRUCTION OF A

## IN GENERAL

- GIVEN DEGREE SEQUENCE

$$d = (d_1, \dots, d_N)$$

- CONSTRUCT MULTIGRAPH A  
WITH N NODES, DEG. SEQUENCE d  
SUCH THAT IS(A) "IS SMALL"



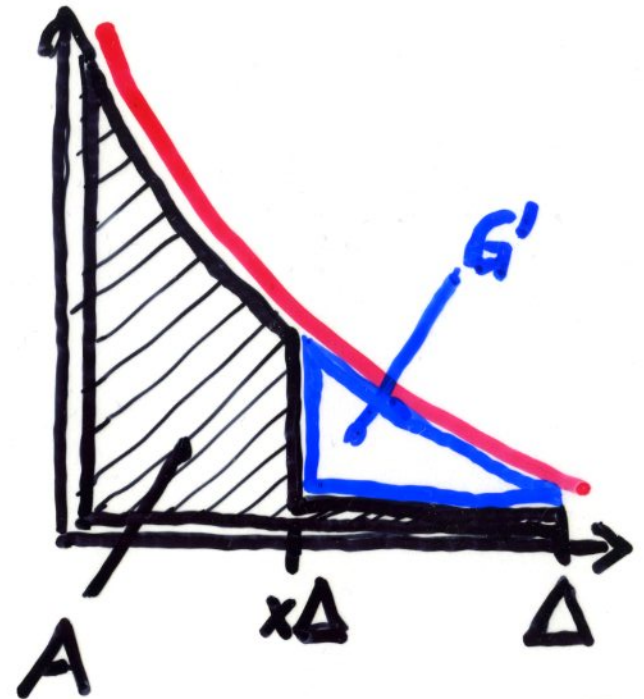
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## IN GENERAL

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## APPROACH



# CONSTRUCTION OF A

► DEGREE SEQUENCE FOR  
INTERVAL  $[a, b]$ ,  $1 \leq a \leq b \leq \Delta$

$$d = (\underbrace{a, \dots, a}_{\substack{e^a \\ a^B}}, a+1, \dots, a+1, \dots, \underbrace{b, \dots, b}_{\substack{e^b \\ b^B}})$$

# CONSTRUCTION OF $A$

- ▶ DEGREE SEQUENCE FOR INTERVAL  $[a, b]$ ,  $1 \leq a \leq b \leq \Delta$

$$d = (\underbrace{a, \dots, a}_{\frac{e^{\alpha}}{a^{\beta}}}, a+1, \dots, a+1, \dots, \underbrace{b, \dots, b}_{\frac{e^{\alpha}}{b^{\beta}}})$$

- ▶ BOUND ON SIZE OF MIS IN  $A$

# CONSTRUCTION OF A

- ▶ DEGREE SEQUENCE FOR INTERVAL  $[a, b]$ ,  $1 \leq a \leq b \leq \Delta$

$$d = (\underbrace{a_1, \dots, a_1}_{\frac{e^\alpha}{a^\beta}}, a_{t+1}, \dots, a_{t+1}, \dots, \underbrace{b_1, \dots, b_1}_{\frac{e^\alpha}{b^\beta}})$$

- ▶ BOUND ON SIZE OF MIS IN A PARTITION OF  $[a, b]$  INTO SUBINTERVALS



# ESTIMATE FOR IS(A)



# ESTIMATE FOR IS(A)



$$IS(A) = IS(G_{[a,b]}) \leq \sum_{j=0}^{k-1} \left[ \frac{\sum_{l=a_j}^{a_{j+1}-1} \lfloor \frac{e^x}{l^B} \rfloor}{a_j} \right]$$



# ESTIMATE FOR IS(A)



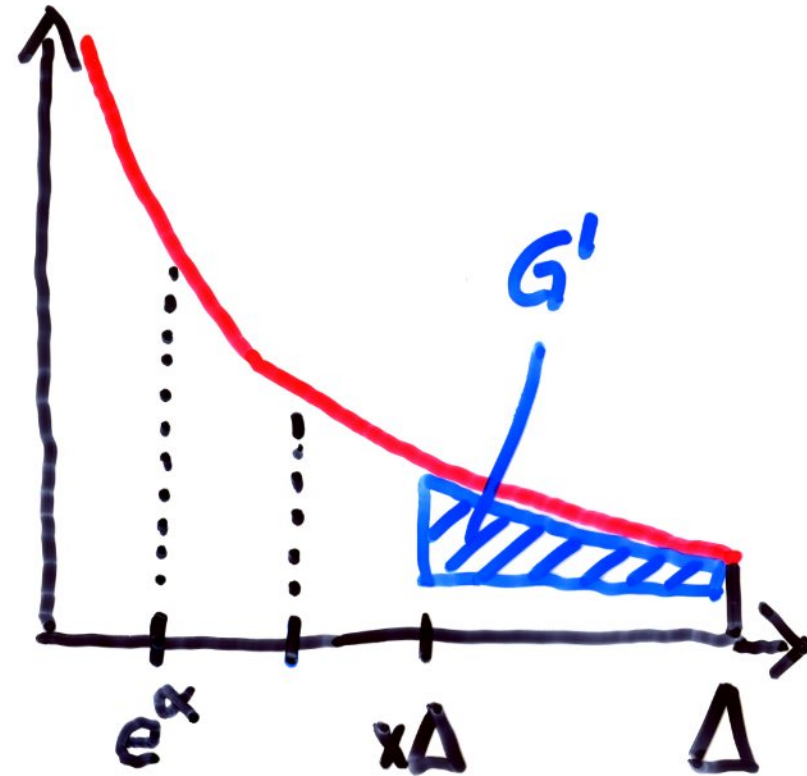
$$IS(A) = IS(G_{[a,b]}) \leq \sum_{j=0}^{k-1} \left[ \frac{\sum_{l=a_j}^{a_{j+1}-1} \left\lfloor \frac{e^x}{l^\beta} \right\rfloor}{a_j} \right]$$

▶ FOR  $\beta < 1$  :  $k = \text{CONST}$

▶ FOR  $\beta = 1$  :  $k = \log n$

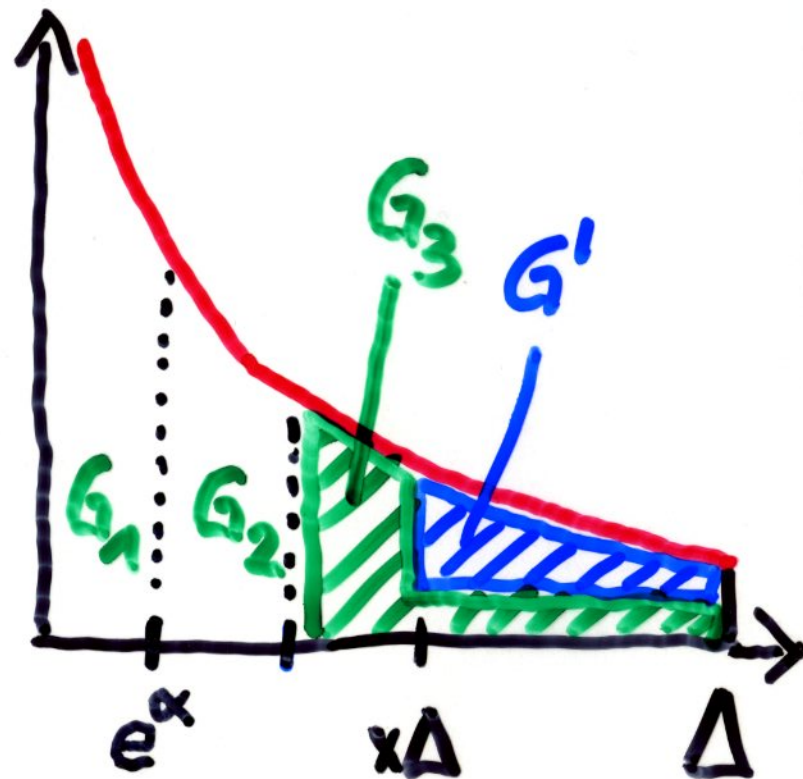
# THE CASE $\beta < 1$ CONT.

► SPLIT  $[1, \Delta]$  INTO FOUR PARTS



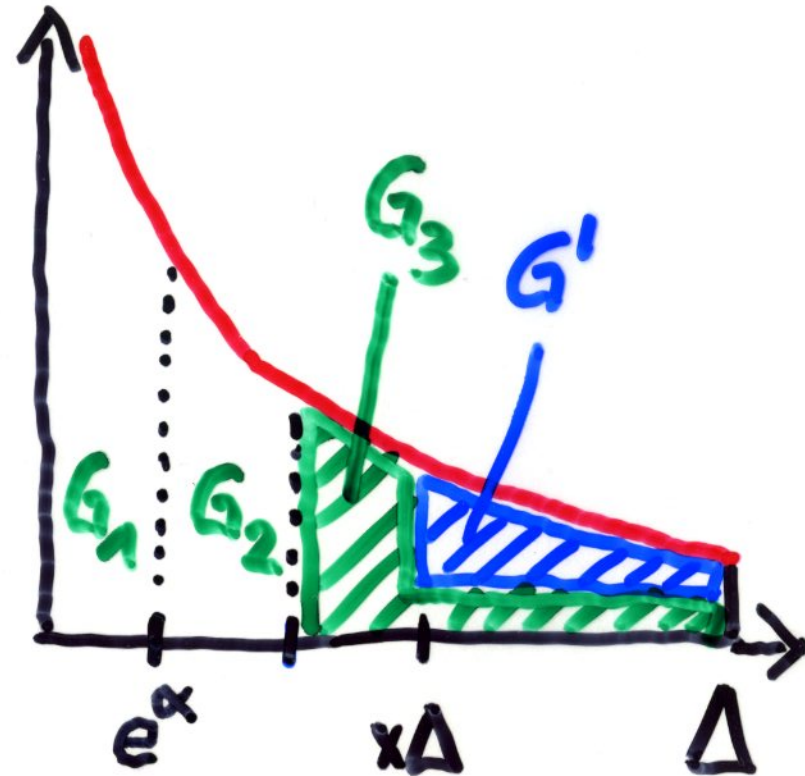
# THE CASE $\beta < 1$ CONT.

- ▶ SPLIT  $[1, \Delta]$  INTO FOUR PARTS
- ▶ FOR  $i=1, 2, 3$  SHOW THAT  
$$IS(G_i) \leq \text{CONST} \cdot e^\alpha$$



## THE CASE $\beta < 1$ CONT.

- ▶ SPLIT  $[1, \Delta]$  INTO FOUR PARTS
- ▶ FOR  $i=1,2,3$  SHOW THAT  
$$IS(G_i) \leq \text{CONST} \cdot e^\alpha$$



### THEOREM

FOR  $\beta < 1$ , MIS ON  $(\alpha, \beta)$ -PLG  
IS NP-HARD TO APPROXIMATE  
WITHIN  $n^{1-\beta-\epsilon}$

## THE CASE $\beta=1$

WE HAVE

$$\# \text{ NODES} = \alpha e^{\alpha}$$

$e^{\alpha}$  NODES OF DEGREE 1

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$e^{\alpha}$  NODES OF DEGREE 1

$\Rightarrow O(\log n)$ -APPROX. FOR MIS  
IN  $(\alpha, \beta)$ -PLG FOR  $\beta=1$

# THE CASE $\beta=1$

WE HAVE

$$\# \text{ NODES} = \alpha e^{\alpha}$$

$e^{\alpha}$  NODES OF DEGREE 1

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QUESTION APPROXIMATION LOWER BOUND?

## THE CASE $\beta = 1$

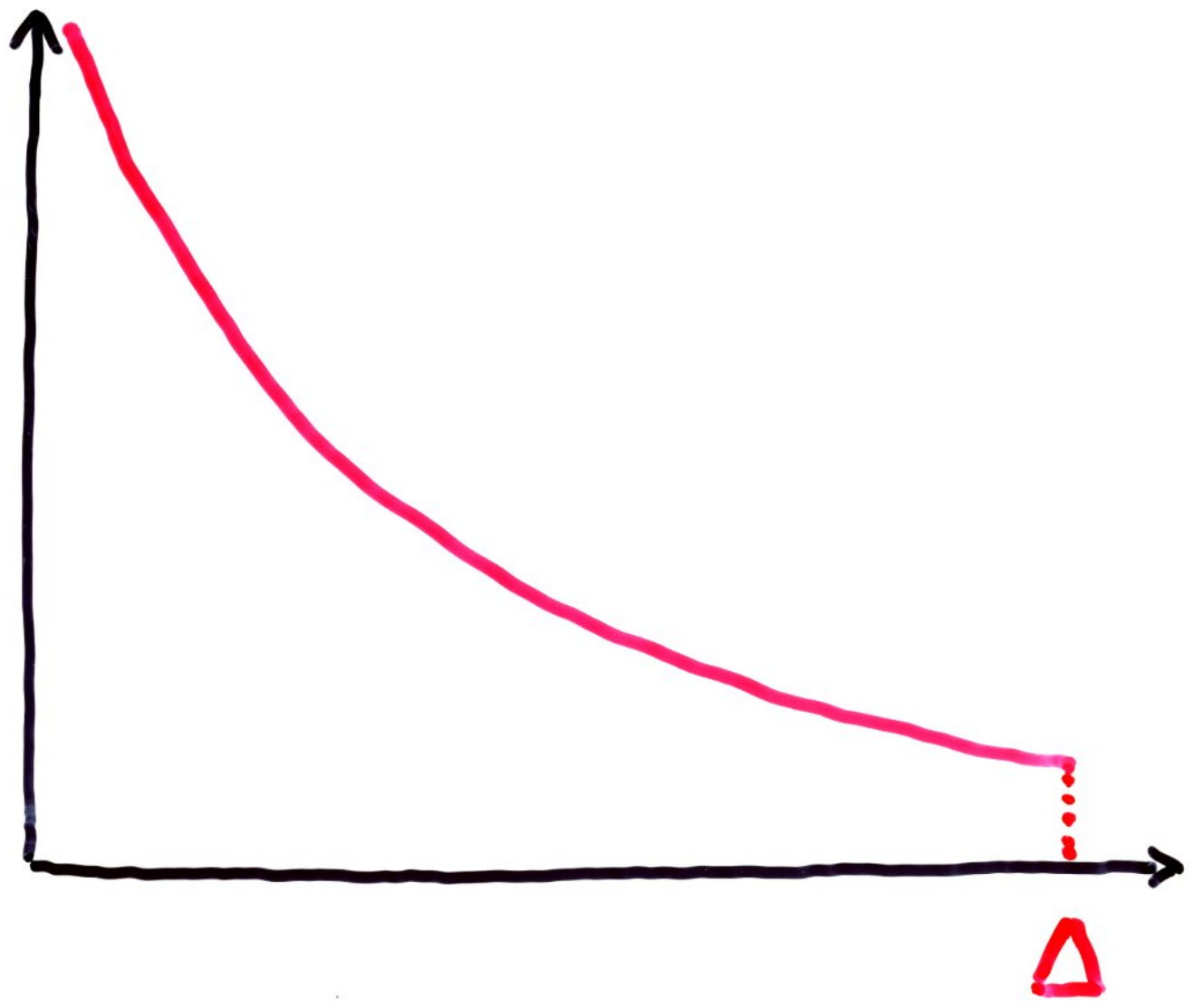
- START FROM ALON ET AL. '95  
LOWER BOUND  $\Delta^\epsilon$  FOR MIS  
IN GRAPHS WITH MAX. DEGREE  $\leq \Delta$



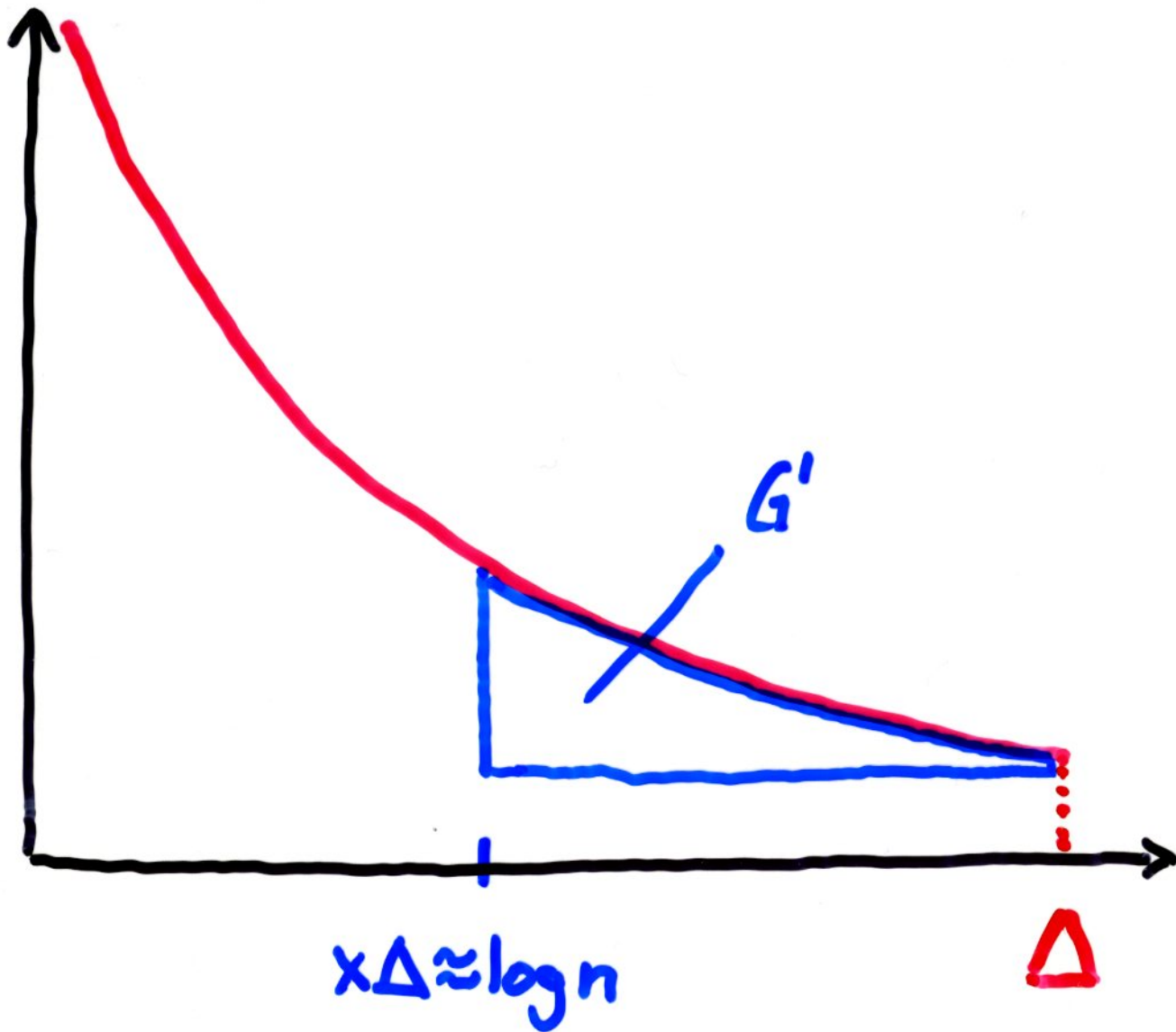
## THE CASE $\beta = 1$

- START FROM ALON ET AL. '95  
LOWER BOUND  $\Delta^\epsilon$  FOR MIS  
IN GRAPHS WITH MAX. DEGREE  $\leq \Delta$
- SHOW: THIS ALSO HOLDS FOR  $\Delta = \log n$

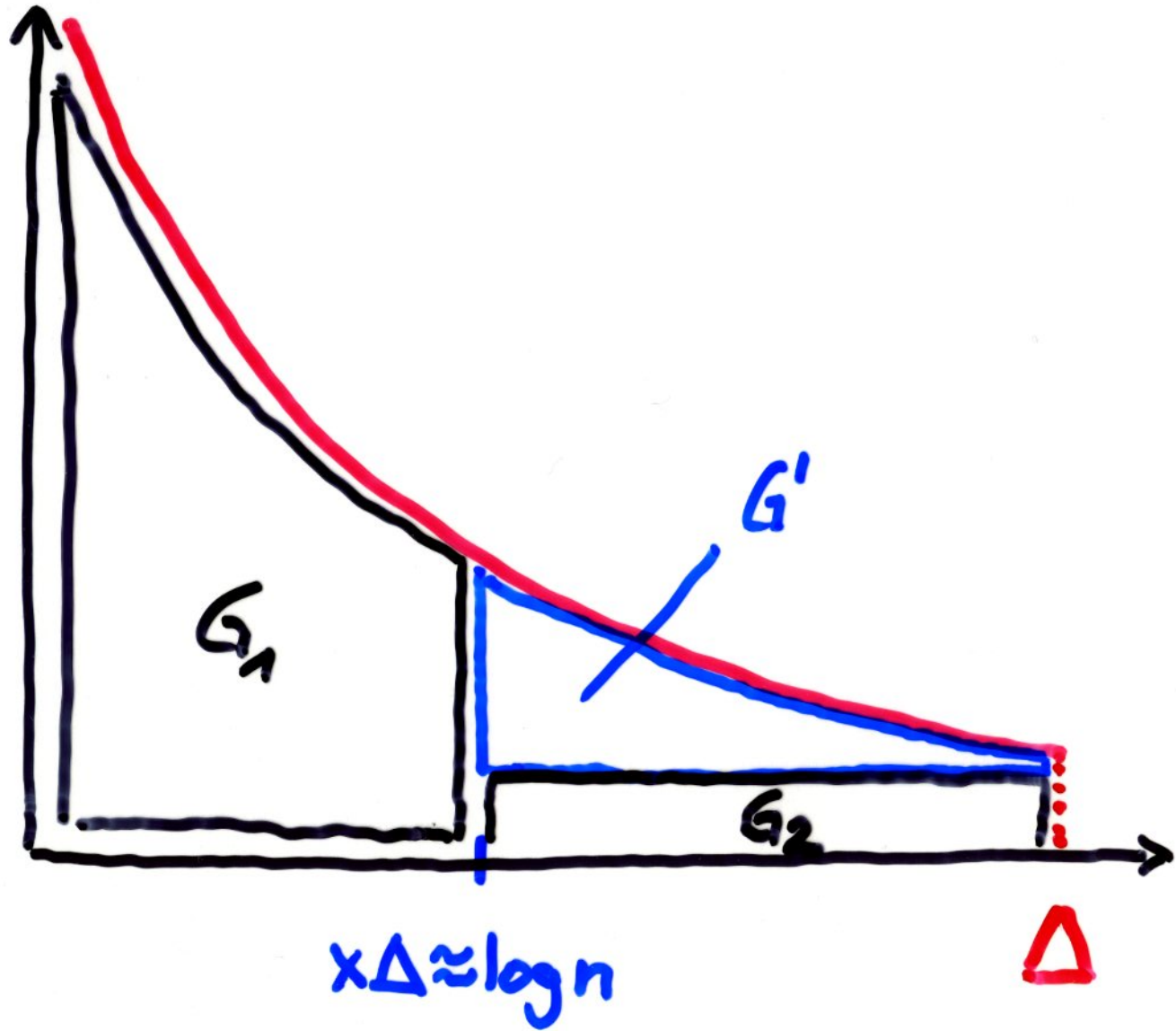
$\beta = 1$  : THE EMBEDDING



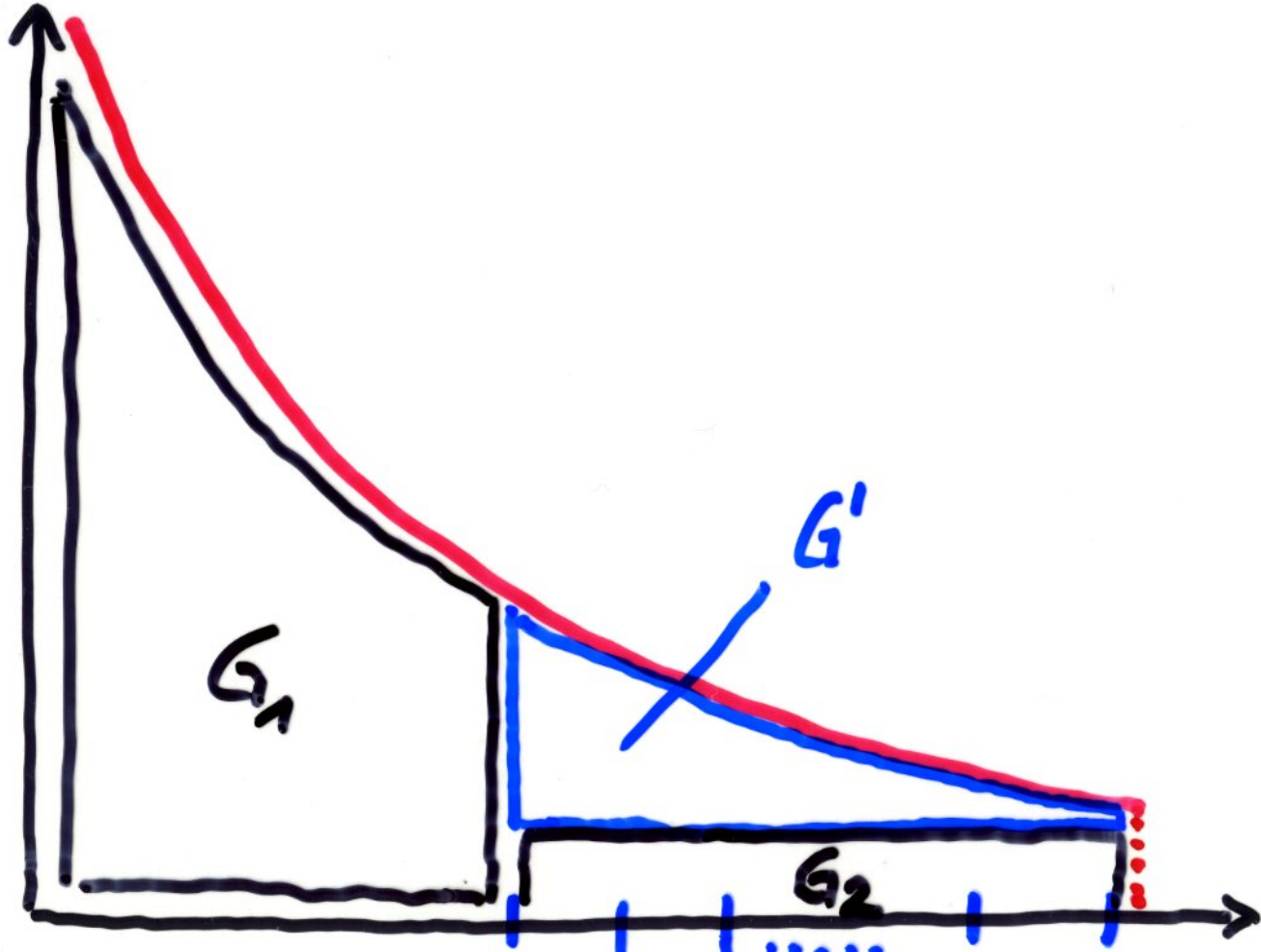
$\beta = 1$ : THE EMBEDDING



$\beta = 1$ : THE EMBEDDING



$\beta = 1$ : THE EMBEDDING



$x\Delta \approx \log n$

▶ SPLIT INTO  $\alpha = \log n$  PARTS

# THE CASE $\beta=1$



THEOREM  
FOR  $\beta=1$ , THERE EXISTS  $\epsilon > 0$   
SUCH THAT MIS IN  $(\alpha, \beta)$ -PLG  
IS NP-HARD TO APPROX. WITHIN  $(\log n)^\epsilon$ .

# OPEN PROBLEMS

- LOWER BOUND  $(\log n)^{1-\epsilon}$  FOR  $\beta=1$  ?

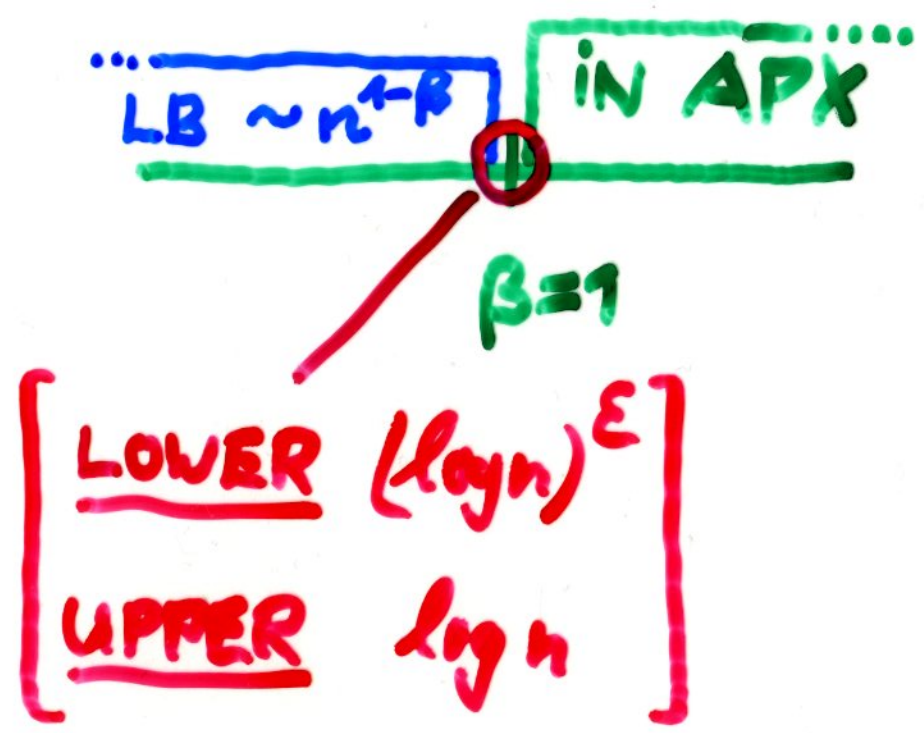
# OPEN PROBLEMS

- LOWER BOUND  $(\log n)^{1-\epsilon}$  FOR  $\beta=1$  ?
- FUNCTIONAL CASES ?



# OPEN PROBLEMS

- LOWER BOUND  $(\log n)^{1-\epsilon}$  FOR  $\beta=1$  ?
- FUNCTIONAL CASES ?



# OPEN PROBLEMS

● LOWER BOUND  $(\log n)^{1-\epsilon}$  FOR  $\beta=1$  ?

● FUNCTIONAL CASES ?

...  $LB \sim n^{\beta}$  | IN APX ...

$\beta=1$

[ LOWER  $(\log n)^{\epsilon}$   
UPPER  $\log n$  ]



WHAT ABOUT

$$\beta = 1 - \frac{1}{f(n)}$$

$$\beta = 1 + \frac{1}{f(n)} ?$$

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