

# A Naive Algorithm for Feedback Vertex Set

Yixin Cao (操宜新)

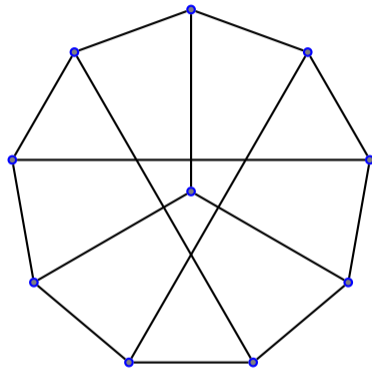
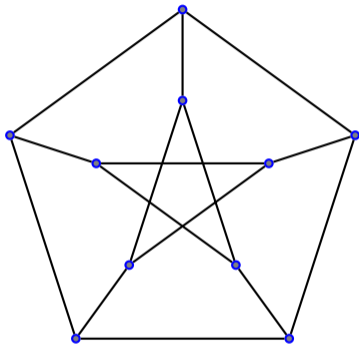
Department of Computing, Hong Kong Polytechnic University

香港理工大學 電子計算學系

Symposium on Simplicity in Algorithms

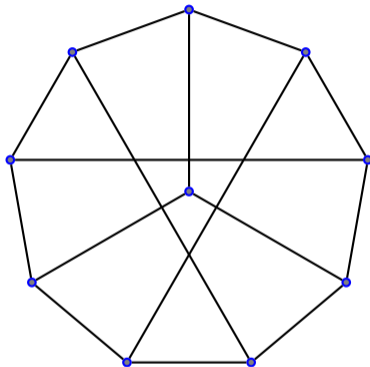
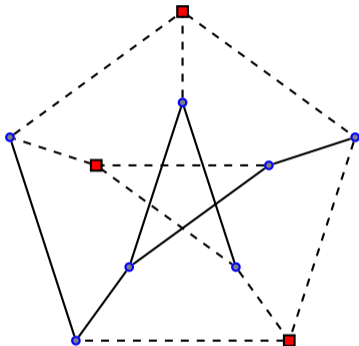
New Orleans, LA January 10, 2018

# The problem



Given a graph  $G$  and an integer  $k$ , the feedback vertex set problem asks for the deletion of at most  $k$  vertices to make  $G$  a forest.

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The constraint satisfaction problem (CSP) asks for the assignment of values to variables to satisfy a set of constraints.

[Freuder 1982]: can be solved in P-time when the constraint graph is a forest.

[Dechter Pearl 1987]: one way to solve the constraint satisfaction problem is to

- 1 find first a minimum feedback vertex set  $V_-$  of the constraint graph,
- 2 enumerate all possible assignments on them, and
- 3 then solve the remaining instance.

On an instance  $I$  on  $p$  variables, it takes  $O(p^{|V_-|} \cdot |I|^{O(1)})$  time.

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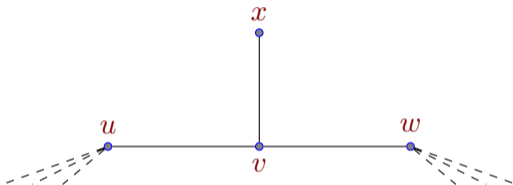
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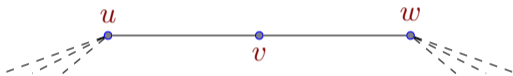


## Low-degree vertices



A vertex of degree 1 can always be safely deleted.

# Low-degree vertices



A degree-2 vertex in a solution can always be replaced with one of its neighbors.

# Low-degree vertices

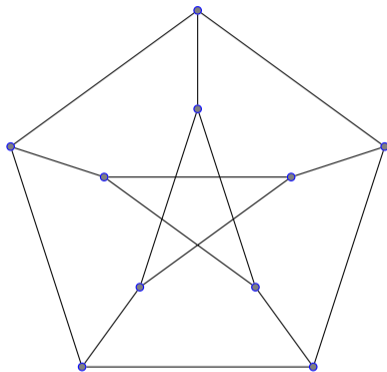


If  $u$  and  $w$  are not adjacent,  
then we can delete  $v$  and add edge  $uw$ .  
this operation is called "smoothen."

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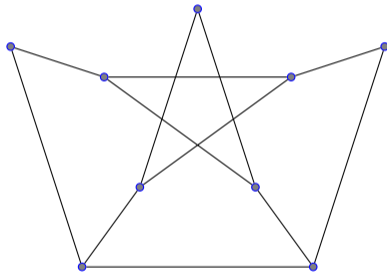
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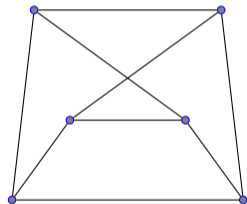
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Greed ... is good. Greed is right. Greed works.

Wall street



- [Erdős Pósa 1962]: A graph of minimum degree  $\geq 3$  has a cycle of length  $O(\log n)$ .
- A trivial  $O(\log n)$  approximation: find a shortest cycle, remove all vertices.
  - A nontrivial  $(\log k)^{O(k)}$ -time parameterized algorithm: after a polynomial kernel.

## Large-degree vertices

[Becker Geiger 1996] [Bafna Berman Fujito 1999]: A greedy 2-approximation (Local ratio)

[Chudak Goemans Hochbaum Williamson 1998]: A primal-dual interpretation

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# Parameterized algorithms

Algorithms running in time  $O(f(k) \cdot n^{O(1)})$ , where  $k$  is a parameter (the solution size).

	$f(k)$
Downey Fellows 1992	$(2k + 1)^k$
Bodlaender 1994	$17(k^4)!$
Raman et al. 2002	$\max\{12^k, (4 \log k)^k\}$
Kanj et al. 2004	$(2 \log k + 2 \log \log k + 18)^k$
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Dehne et al. 2005	$10.6^k$
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All  $c^k$  algorithms use technique  
“iterative compression.”

	vertex cover	feedback vertex set
to kill	edges	cycles
to make	independent set (edgeless)	forest (acyclic)
	treewidth 0	treewidth $\leq 1$
approx	2	2
parameterized	$1.2738^k$	$3.62^k$

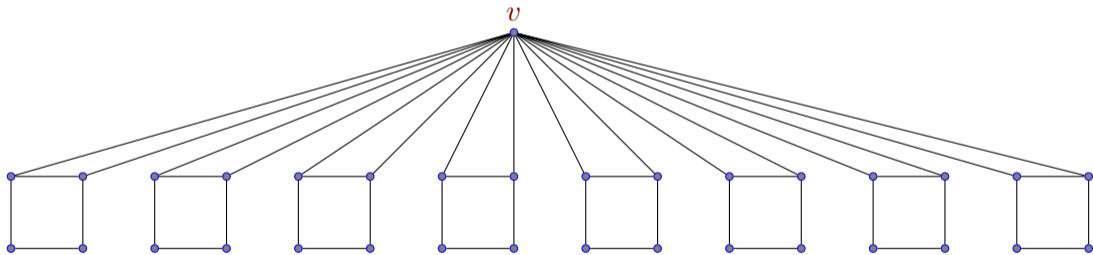
the *simplest nontrivial* vertex deletion problem

# The Algorithm



A vertex of the largest degree is highly likely in the solution

A vertex of the largest degree is highly likely in the solution, but *not* always.



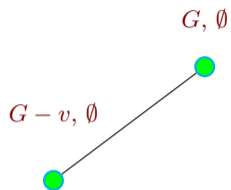
# A branching algorithm

$G, \emptyset$



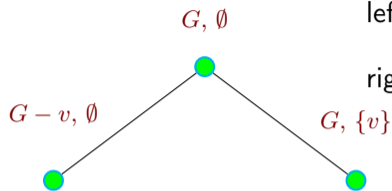
each node has two parts,  
the graph  $G$  and the set  $F$  of *permanent* vertices.

# A branching algorithm



left child: the vertex is put into the solution,  
hence deleted from  $G$

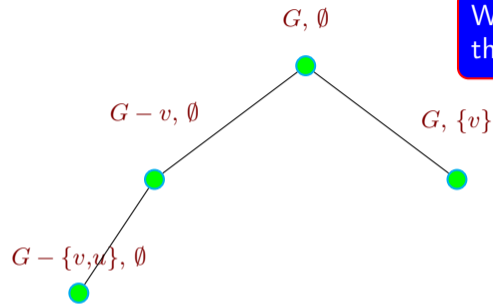
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left child: the vertex is put into the solution,  
hence deleted from  $G$

right child: the vertex is not in the solution,  
hence put into  $F$

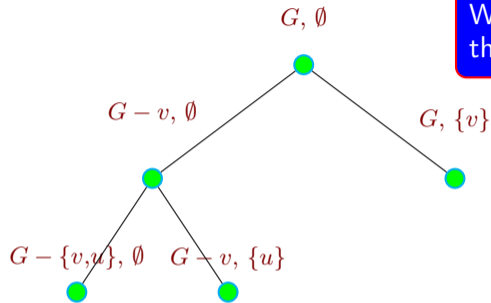
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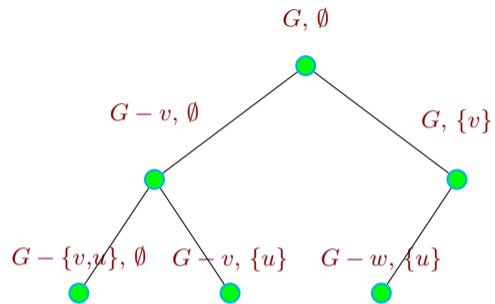
We always choose the vertex with the largest degree in  $V(G) \setminus F$  to branch.

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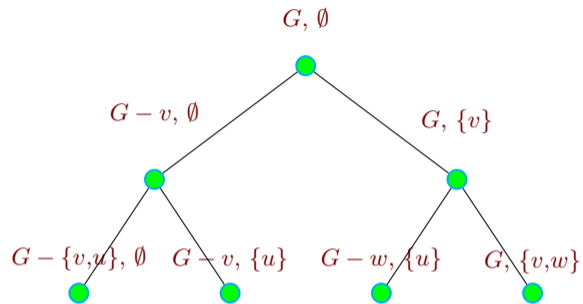


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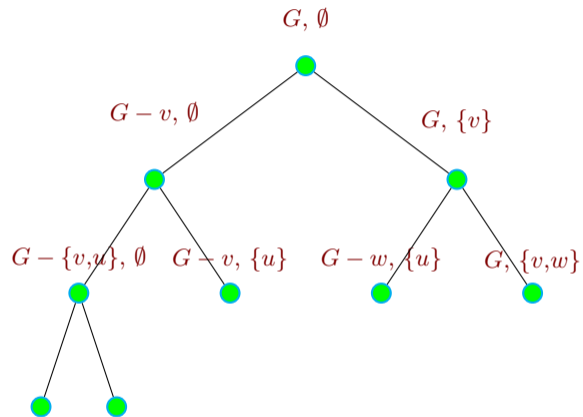




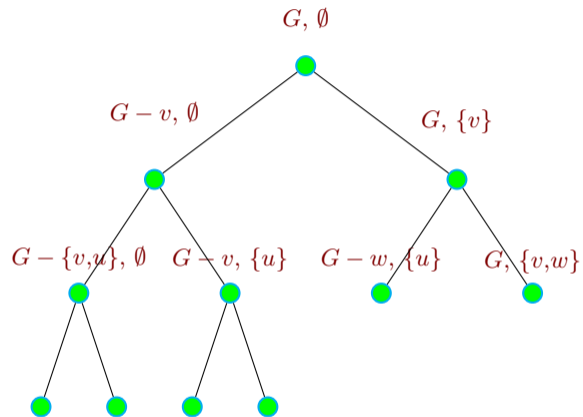
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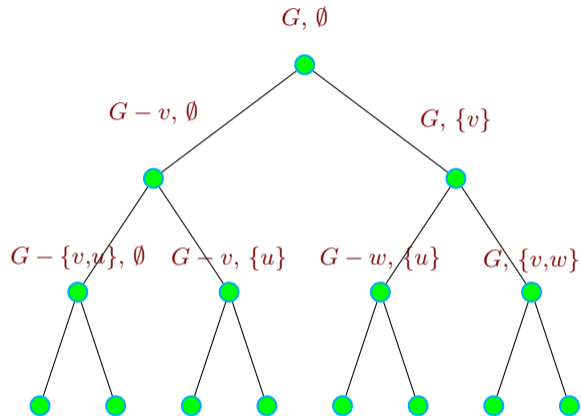
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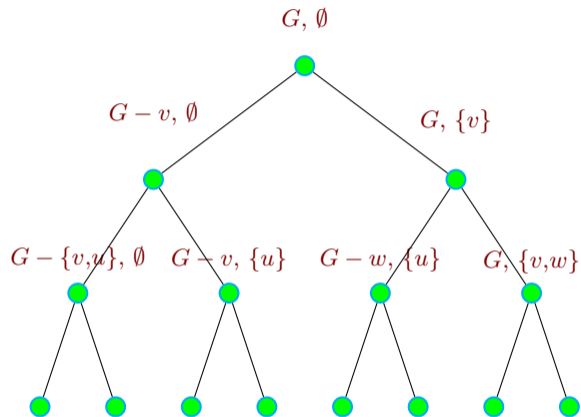


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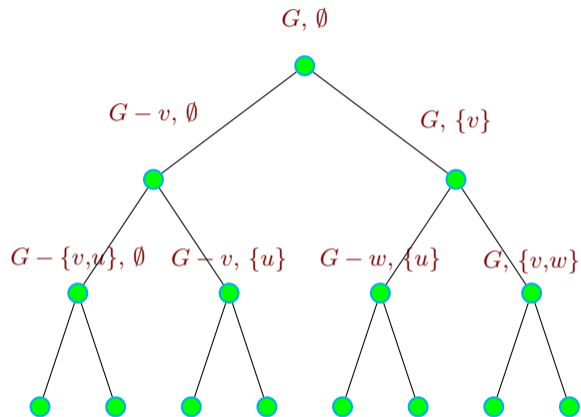
The key of the analysis is to bound the number of nodes, which boils down to bounding the depth.

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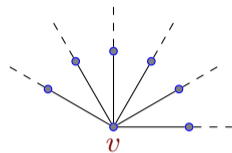
We cannot go left more than  $k$  times.  
We're hence focused on the right steps.

# A branching algorithm



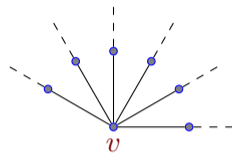
Let's fix an execution path.

## The Analysis



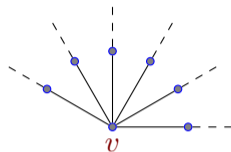


$$\sum_{v \in V(T)} d(v) = 2|E(T)| = 2|V(T)| - 2$$



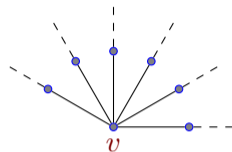
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$L$ : leaves;  $V_3$ : vertices of degree  $\geq 3$ .



# Elementary facts of trees

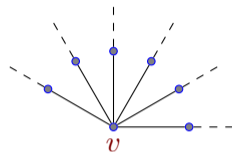
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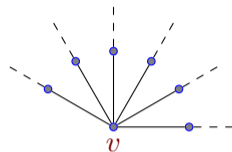
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For each  $v \notin V_3$ ,

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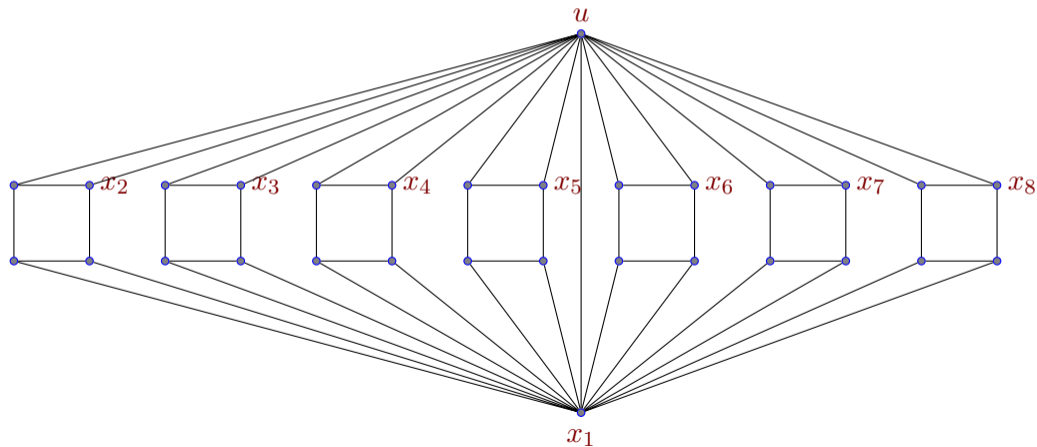
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This correlates deleted vertices and permanent vertices.

# The example

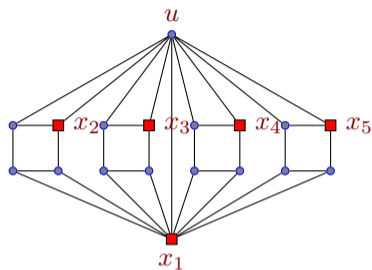
the first vertex  $x_1$  is put into  $V_-$ , second  $u$  into  $F$ ,  
then the deletion of each  $x_i$  decreases its degree by two.



# Algorithm invariants

- ① During the algorithm, the degree of no vertex can increase.
- ② There is no vertex of degree 0 or 1 in the graph when a recursive call is made.

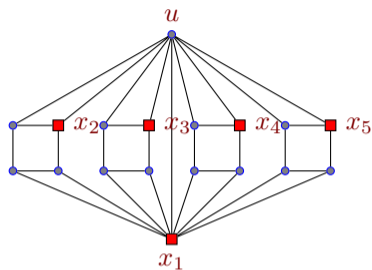
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- $d^*(v)$ : the degree of  $v$  at the moment it is put into  $V_-$  or  $F$ .  $d(v) \geq d^*(v)$
- effective decrements of the degree of a vertex  $u \in F$ : from  $d^*(u)$  to 2.
- an effective decrement is incurred by  $x_i \in V_-$  if it is after deleting  $x_i$ .  
 $\delta(u, x_i)$ : #effective decrements of  $u \in F$  incurred by  $x_i \in V_-$ .
- $\delta(u, x_i)$  may be larger than 1.
- $\delta(u, x_i) > 0 \Rightarrow u \in F$  when  $x_i$  is deleted..



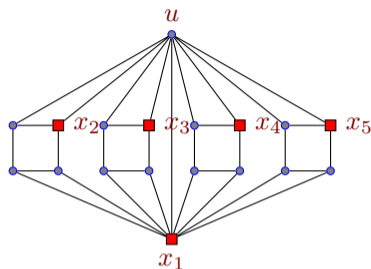
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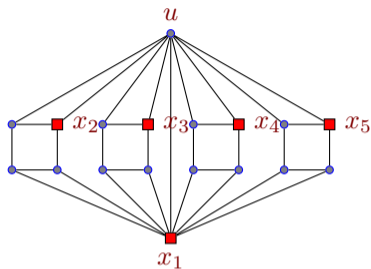


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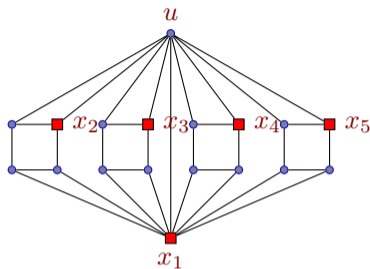
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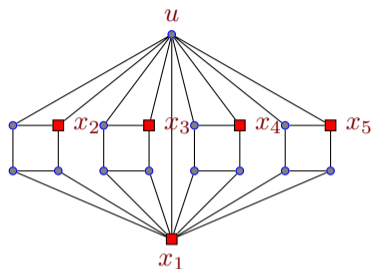
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decision points:  $x_1 \rightarrow u \rightarrow x_2 \rightarrow x_3 \rightarrow x_4 \rightarrow x_5$ .

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$\delta(u, x_1) = 0$ ;  $\delta(u, x_2) = \delta(u, x_3) = \delta(u, x_4) = 2$ ;  $\delta(u, x_5) = 0$

- $d^*(v)$ : the degree of  $v$  at the moment it is put into  $V_-$  or  $F$ .  $d(v) \geq d^*(v)$
- effective decrements of the degree of a vertex  $u \in F$ : from  $d^*(u)$  to 2.
- an effective decrement is incurred by  $x_i \in V_-$  if it is after deleting  $x_i$ .  
 $\delta(u, x_i)$ : #effective decrements of  $u \in F$  incurred by  $x_i \in V_-$ .
- $\delta(u, x_i)$  may be larger than 1.
- $\delta(u, x_i) > 0 \Rightarrow u \in F$  when  $x_i$  is deleted..

Lemma 1: For any  $u \in F$  and  $v \in V_-$ , if  $\delta(u, v) > 0$  then  $d^*(u) \geq d^*(v)$ .

Lemma 2:  $\sum_{u \in F} \delta(u, v) \leq d^*(v)$  for each  $v \in V_-$ .

$$\begin{aligned}
 |V_-| &= \sum_{v \in V_-} 1 = \sum_{v \in V_-} \frac{d^*(v)}{d^*(v)} \geq \sum_{v \in V_-} \frac{1}{d^*(v)} \sum_{u \in F} \delta(u, v) \\
 &= \sum_{v \in V_-} \sum_{u \in F} \frac{\delta(u, v)}{d^*(v)} \geq \sum_{v \in V_-} \sum_{u \in F} \frac{\delta(u, v)}{d^*(u)} = \sum_{u \in F} \frac{1}{d^*(u)} \sum_{v \in V_-} \delta(u, v) \\
 &= \sum_{u \in F} \frac{d^*(u) - 2}{d^*(u)} \geq \sum_{u \in F} \frac{1}{3} = \frac{|F|}{3},
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Lemma 3: Therefore, an execution path leading to a solution has depth at most  $4k$ .

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We terminate all execution paths after  $4k$  steps  $\Rightarrow O(16^k \cdot n^2)$ .  $\square$

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Only put vertices of degree  $\geq 4$  into  $F$ , then

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We terminate all execution paths after  $3k$  steps  $\Rightarrow O(8^k \cdot n^2)$ .  $\square$

Beauty is the first test:  
there is no permanent place in the world for ugly mathematics..

G. H. Hardy

To theorists:  
stop ignoring successful heuristic algorithms by pretending their nonexistence!



thanks!