

Lecture 7 (part II)

Testing dense graphs

- bipartiteness

Adjacency Matrix model

G represented by matrix A
st. can query A in
one step

$$A = \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix} A_{ij}$$

Distance from property P :

def G is ϵ -far from P if must change $> \epsilon \cdot n^2$
entries in A to turn G into member of P

Testing "sparse" properties:

Graph type

max degree

natural representation

notion of distance

Previously

sparse

Now

dense

Bipartiteness:

- Can color nodes red/blue st. no edge monochromatic
- Can partition nodes into (V_1, V_2) st.

$$\nexists \underset{(u,v)}{e} \in E \quad \text{st.} \quad \begin{array}{l} u, v \in V_1 \\ \text{or } u, v \in V_2 \end{array}$$

ϵ -far from bipartite: (definition)

- must remove $> \epsilon \cdot n^2$ edges to make bipartite
- \forall partitions $V = (V_1, V_2)$, $> \epsilon \cdot n^2$ violating edges

Testing Algorithms:

- Testing exact bipartiteness;

- Proposed testing algorithm:

- Pick sample of nodes of size $O\left(\frac{1}{\epsilon^2} \log \frac{1}{\epsilon}\right)$

- Consider induced graph on sample

- If bipartite, output PASS
else output FAIL

A first attempt at a proof?

if G bipartite, induced graph is bipartite, so algorithm passes

if G ε -far from bipartite:

must remove εn^2 edges to make it bipartite

equivalently:

Lets try to use the "partition" defn of bipartiteness:

Algorithm 0

Pick $m = \Theta(?)$ random edge slots & query

\forall partitions (V_1, V_2) :

$\text{violating}_{(V_1, V_2)} \leftarrow \# \text{ violating edges in sample wrt } (V_1, V_2)$

If $\exists (V_1, V_2)$ $\text{violating}_{(V_1, V_2)} > 0$ then output FAIL
else output PASS

Wait! How small should δ be?

Recall: All partitions are bad

But: if any partition "looks" good, the algorithm outputs PASS

Probability any partition "looks" good:

for one partition (V_1, V_2) ,

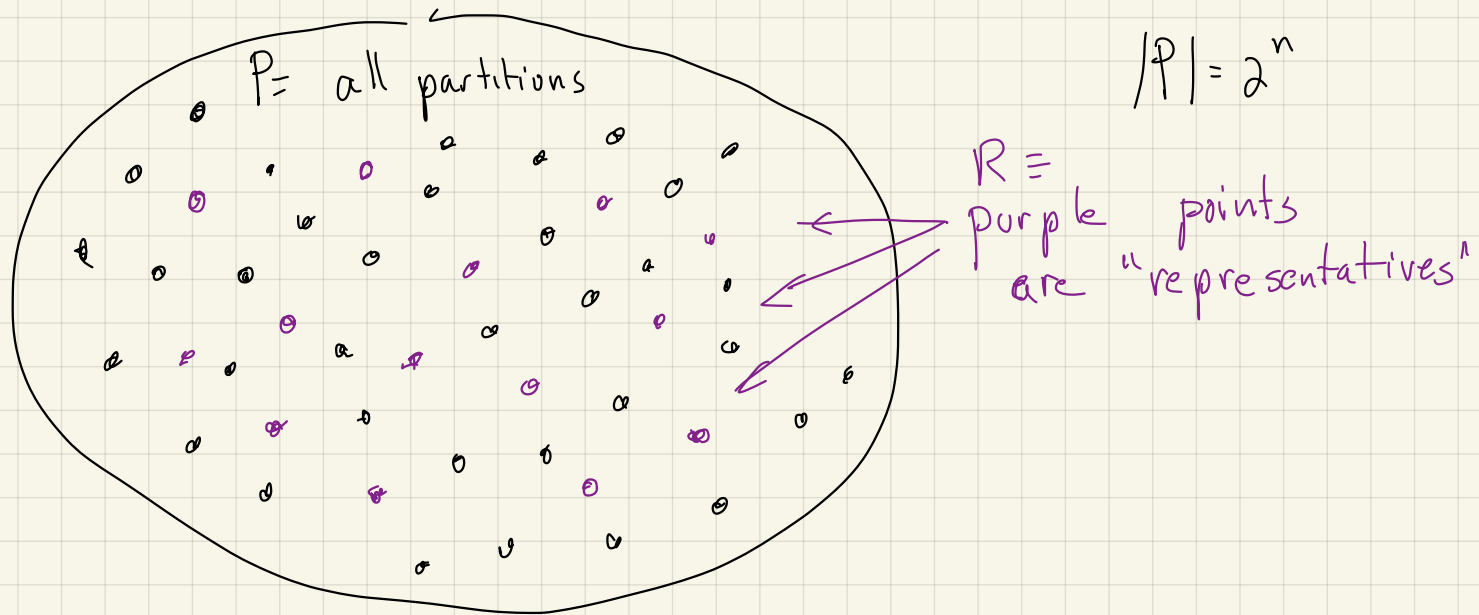
for all partitions (V_1, V_2) ,

Do we really need a union bound?

Do we need to try all partitions?

(Or can we find few "representative" partitions that are close to all partitions?)

Plan: Consider small set of representatives



Useful R satisfies:

$$\forall p \in P \quad \exists r \in R \quad \text{s.t.} \quad \text{dist}(p, r) \leq \epsilon$$

Hope: (1) testing R tells you something about P

(2) $|R| \ll |P|$

(so union bound isn't
as bad)

Plan :

find "representative" partitions s.t.
all partitions are $\frac{\epsilon}{2}$ -close to
some representative.

• if G ϵ -far from bipartite then
 \forall partitions $\geq \epsilon n^2$ violating edges

• if G bipartite then
 \exists partition with 0 violating edges

Algorithm 1

1. pick U, U' randomly from V

" " $\Theta\left(\frac{1}{\epsilon} \log \frac{1}{\epsilon}\right)$ nodes $\Theta\left(\frac{1}{\epsilon} \log \frac{1}{\epsilon}\right)$ nodes

If U not bipartite, FAIL

2. V partitions of U into U_1, U_2 :

• induce partition $(U_1 \cup W_1, U_2 \cup W_2)$ on whole graph:

Partition: $\forall v \in V$ (including $v \in U$)

• if v has nbr in U_2 , put in W_1

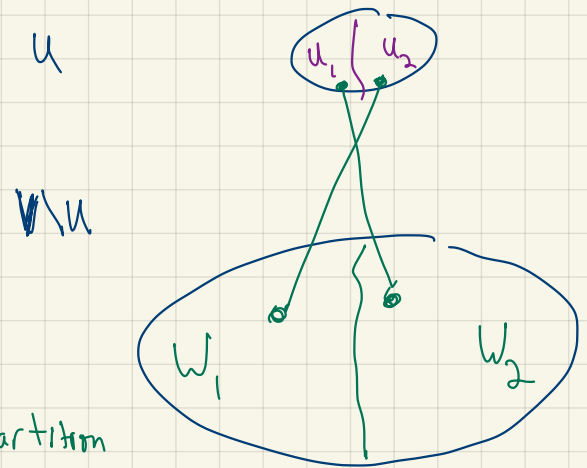
• if v " " " U_1 , " " W_2

• " " " " both \Rightarrow bad partition!
continue to next partition

• " " " " neither, put in W_1

• count $\# \{(u, v) \in E \text{ s.t. violate } [Z_1, Z_2]\}$
if fraction $\leq 3/4 \epsilon$ PASS
else continue to next partition

3. FAIL



Algorithm 1

1. pick U, U' randomly from V

$\Theta(\frac{1}{\epsilon} \log \frac{1}{\epsilon})$ nodes

used to define set of partitions

nodes

used to define random edges:

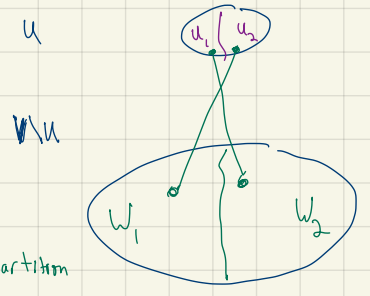
pair off $U' = \{u_1, v_1, u_2, v_2, \dots\}$ to $P = \{(u_1, v_1), (u_2, v_2), \dots\}$ pairs

If U not bipartite, FAIL

2. V partitions of U into U_1, U_2 :

← consider only $2^{|U|}$ partitions. enough?

• induce partition (Z_1, Z_2) on whole graph: U



Compute in time $O(|U|)$ don't do this for all $v \in V$, just those in U'

Partition: $\forall v \in V$ (including $v \in U$)
 • if v has nbr in U_2 , put in W_1
 • if v " " " " U_1 , " " W_2
 • " " " " both \Rightarrow bad partition! continue to next partition
 • " " " " neither, put in W_i

• count # $\{(u, v) \in P \text{ s.t. violate } (Z_1, Z_2)\}$
 if fraction $\leq 3/4 \epsilon$ PASS + half
 else continue to next partition

} why pass if see any violations? we aren't checking all partitions

3. FAIL

Query Complexity:

$O(\frac{1}{\epsilon^2} \log \frac{1}{\epsilon})$

Time Complexity:

$O(2^{\frac{1}{\epsilon} \log \frac{1}{\epsilon}})$

← no dependence on n

can improve dependence on ϵ

Behavior: need to show that if G bipartite, likely to pass + if G ϵ -far from bipartite, likely to fail

if G is ϵ -far:

all partitions Z_1, Z_2 have $> \epsilon \cdot n^2$ violating edges (so all those generated by algorithm do too!)

$\forall Z_1, Z_2 \Pr[\text{fraction of violating edges in } P \text{ is } \leq 3/4 \epsilon n^2] \leq \frac{1}{8 \cdot 2^{|U|}}$

$\Pr[\text{PASS}] = \Pr[\text{any partition of } U \text{ generates partition that passes}] \leq 2^{|U|} \cdot \frac{1}{8 \cdot 2^{|U|}} = \frac{1}{8}$



Algorithm 1

1. pick U, U' randomly from V

$\Theta(\frac{1}{\epsilon} \log \frac{1}{\epsilon})$ nodes
 $\Theta(\frac{1}{2\epsilon} \log \frac{1}{\epsilon})$ nodes

used to define set of partitions

used to define random edges:

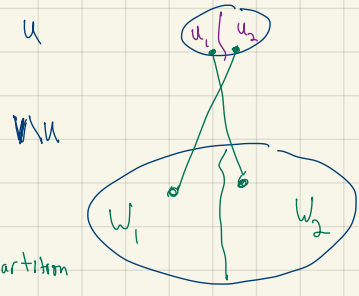
pair off $U' = \{u_1, v_1, u_2, v_2, \dots\}$
 to $P = \{(u_1, v_1), (u_2, v_2), \dots\}$ pairs

If U not bipartite, FAIL

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← consider only $2^{|U|}$ partitions. enough?

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Compute in time $O(|U|)$ → don't do this for all $v \in V$, just those in U

- Partition: $\forall v \in V$ (including $v \in U$)
- if v has nbr in U_2 , put in W_1
 - if v " " " " U_1 , " " W_2
 - " " " " both \Rightarrow bad partition! continue to next partition
 - " " " " neither, put in W_i

• count $\#\{(u,v) \in P \text{ s.t. violate } (Z_1, Z_2)\}$
 if fraction $\leq 3/4 \epsilon$ PASS + half partition
 else continue to next

} why pass if see any violations? we aren't checking all partitions

3. FAIL

if G is bipartite:

does it pass?

Let (Y_1, Y_2) be bipartite partition.

violating edges = 0

For sample U , partition according to Y_1, Y_2 :

$U_1 \leftarrow U \cap Y_1$
 $U_2 \leftarrow U \cap Y_2$

} partition of U

Comments

Can improve runtime to $\text{poly}(1/\epsilon)$

Proposed testing algorithm actually works

In adjacency list model (sparse graphs) need

$\Omega(\sqrt{n})$ queries,

Why more?

Finer grain distinction

dense model: bipartite vs. $\epsilon \cdot n^2$ edges need to be removed

sparse model: bipartite vs. $\epsilon \cdot \Delta \cdot n$ edges need to be r