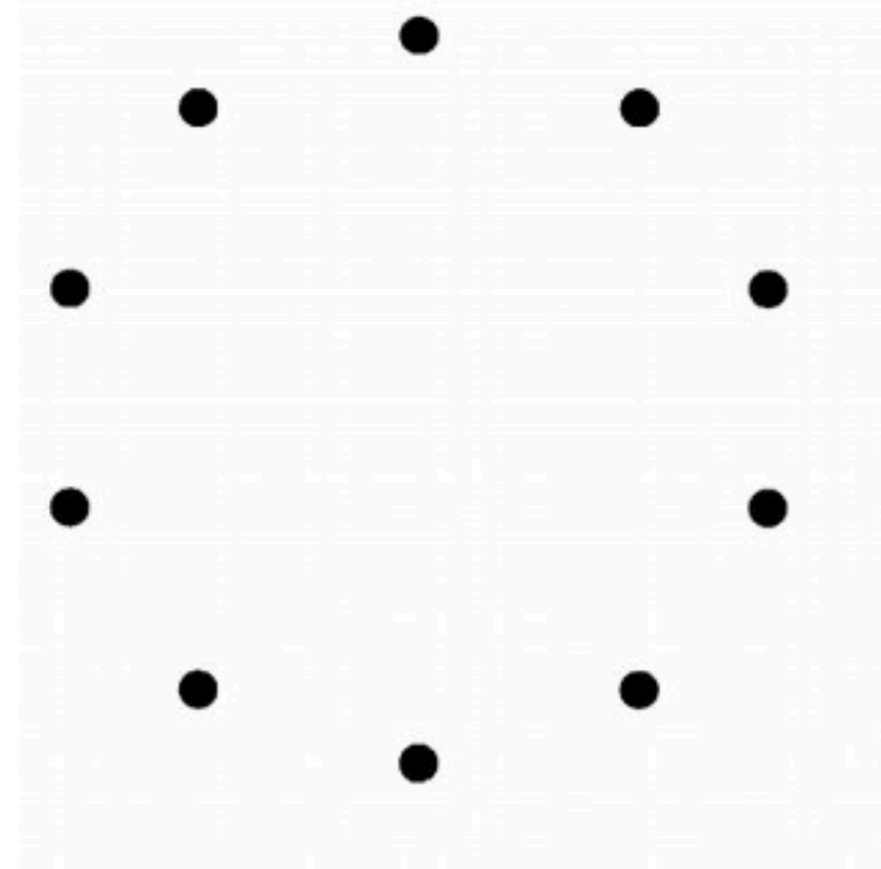


# Robust Estimation for Erdős- Rényi Graphs

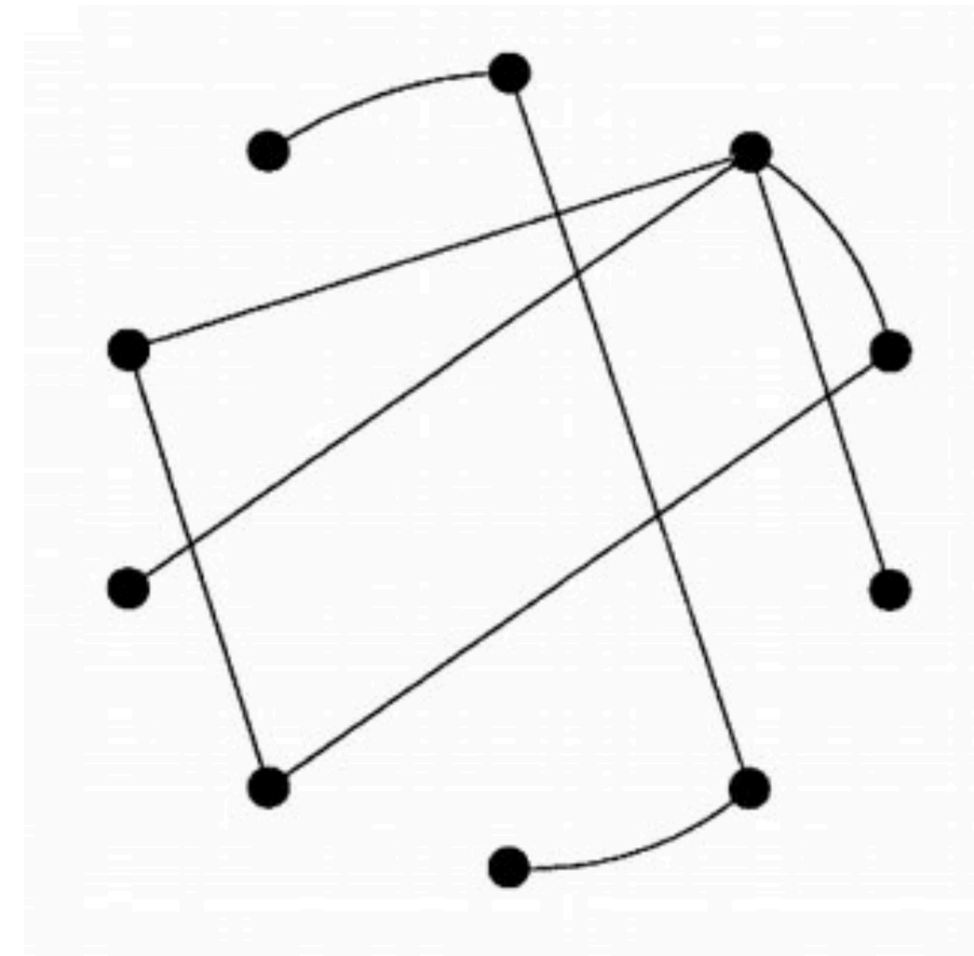
Arjan Chakravarthy, George Chemmala, Heon Lee

# *Erdős-Rényi Random Graph*

- Parameters:  $n, p$
- $n$  nodes



# *Erdős-Rényi Random Graph*



- Parameters:  $n, p$
- $n$  nodes
- Each edge independently has probability  $p$  of existing

# *Motivation*

- Natural / baseline model for real world phenomena
  - social networks
  - epidemiology

Can we estimate  $p$  given that an  $\varepsilon$ -fraction of the nodes have been corrupted?

# Adversarial Model

- $\varepsilon$ -**omniscient** adversary:
  - observes the realization of graph  $G \sim G(n, p)$
  - choose  $B \subset V$  and rewire its edges
- $\varepsilon$ -**oblivious** adversary:
  - choose  $B \subset V$  and distribution on  $B$  without observing the realization of  $G$
- $(q, \varepsilon)$ -omniscient / oblivious adversary

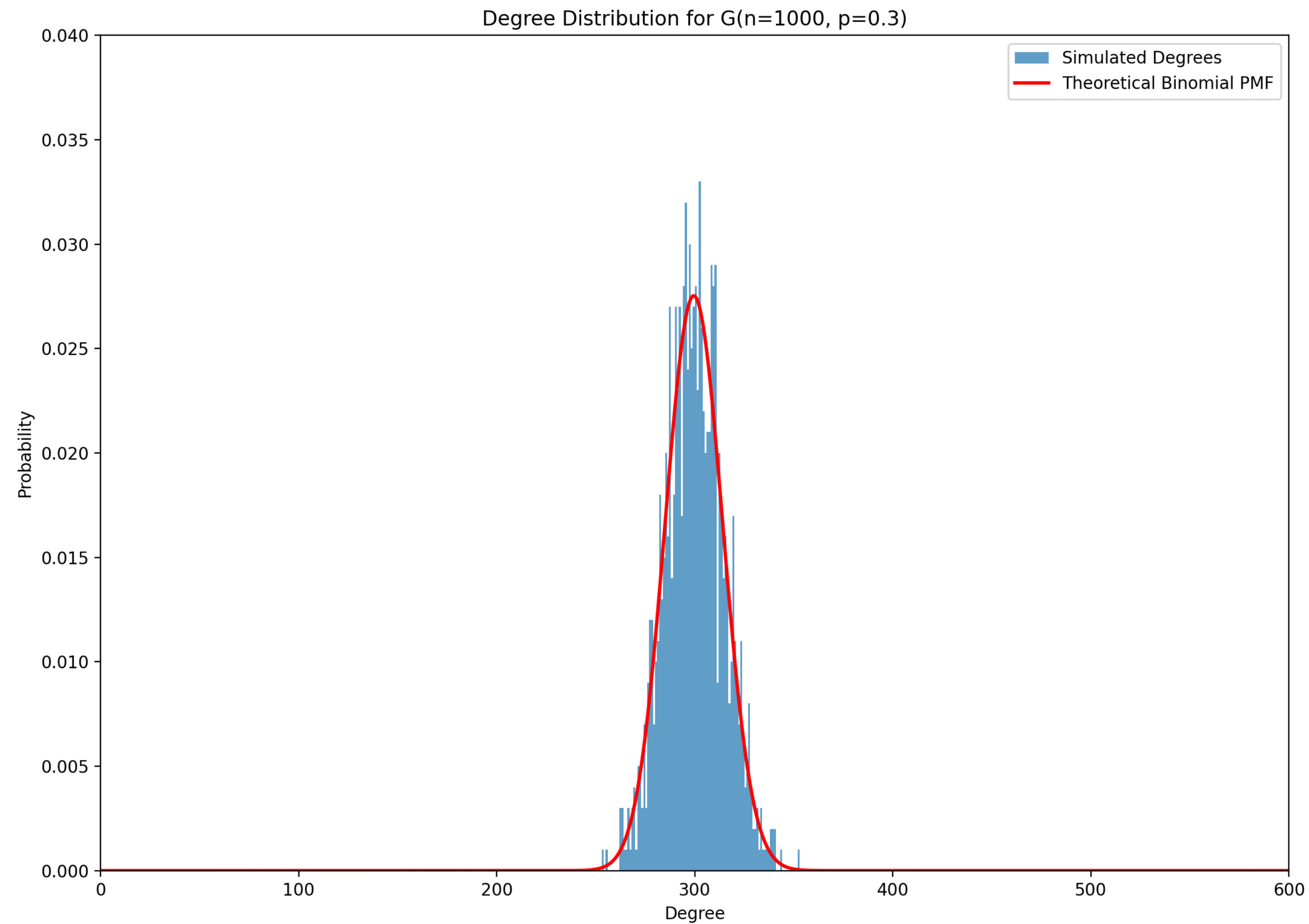
# Existing Algorithms

Methods	Error	Runtime	Authors
Mean/Median	$O(\varepsilon)$	$O(n)$	Acharya et al.
Prune then Mean/Median	$O(\varepsilon^2)$	$O(n \log(\epsilon n))$	Acharya et al.
Spectral Method	$O\left(\frac{\sqrt{p(1-p)\log n}}{n} + \frac{\varepsilon\sqrt{p(1-p)\log(1/\varepsilon)}}{\sqrt{n}} + \frac{\varepsilon}{n}\log n\right)$	$O\left(\varepsilon n^3 \frac{\log n}{\gamma}\right)$	Acharya et al.
Median – Mean	?	?	Us
Variance Method	?	?	Us

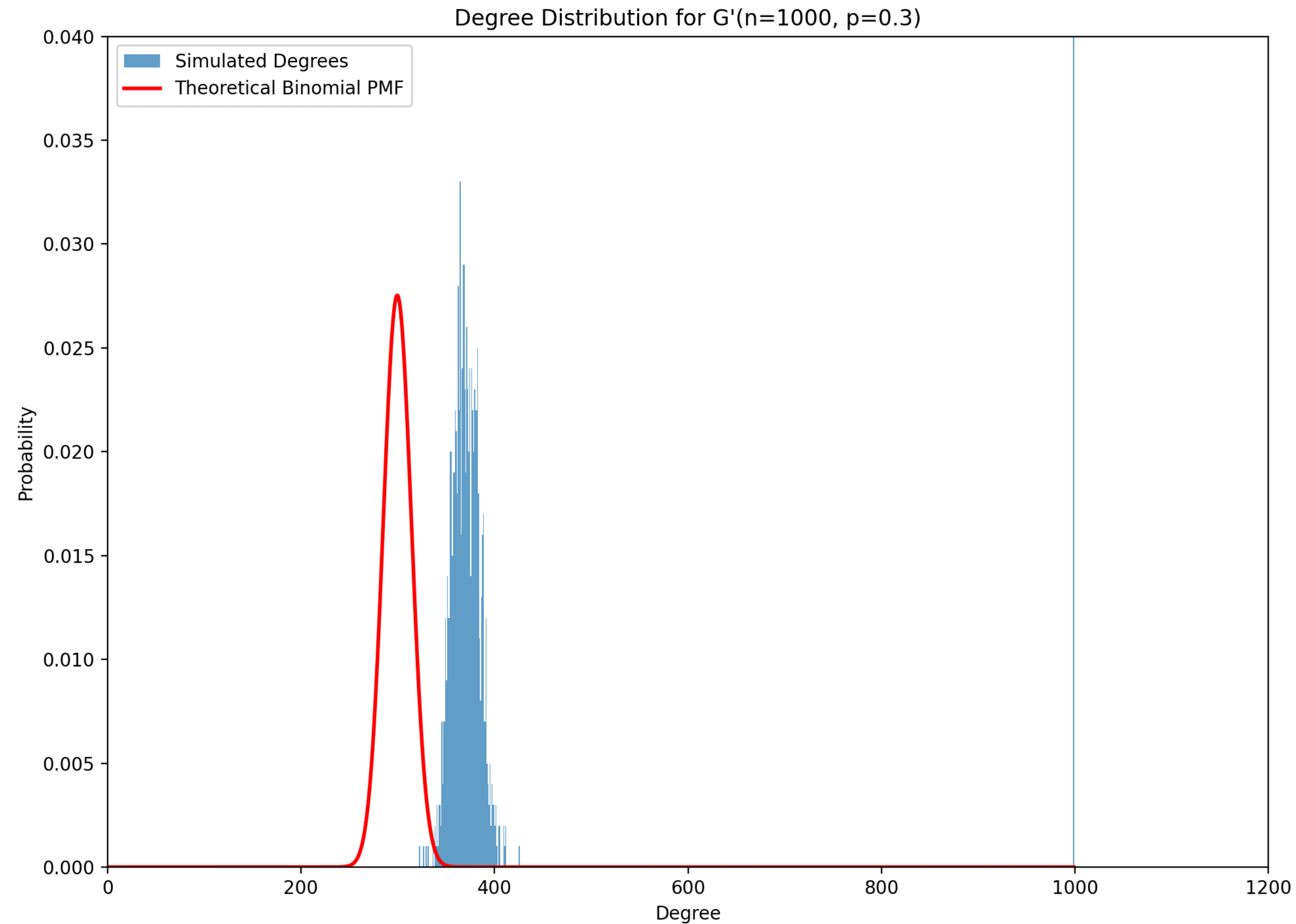
Mean / Median



# *Degrees of Unperturbed Graph*



# *Degrees of Perturbed Graph*

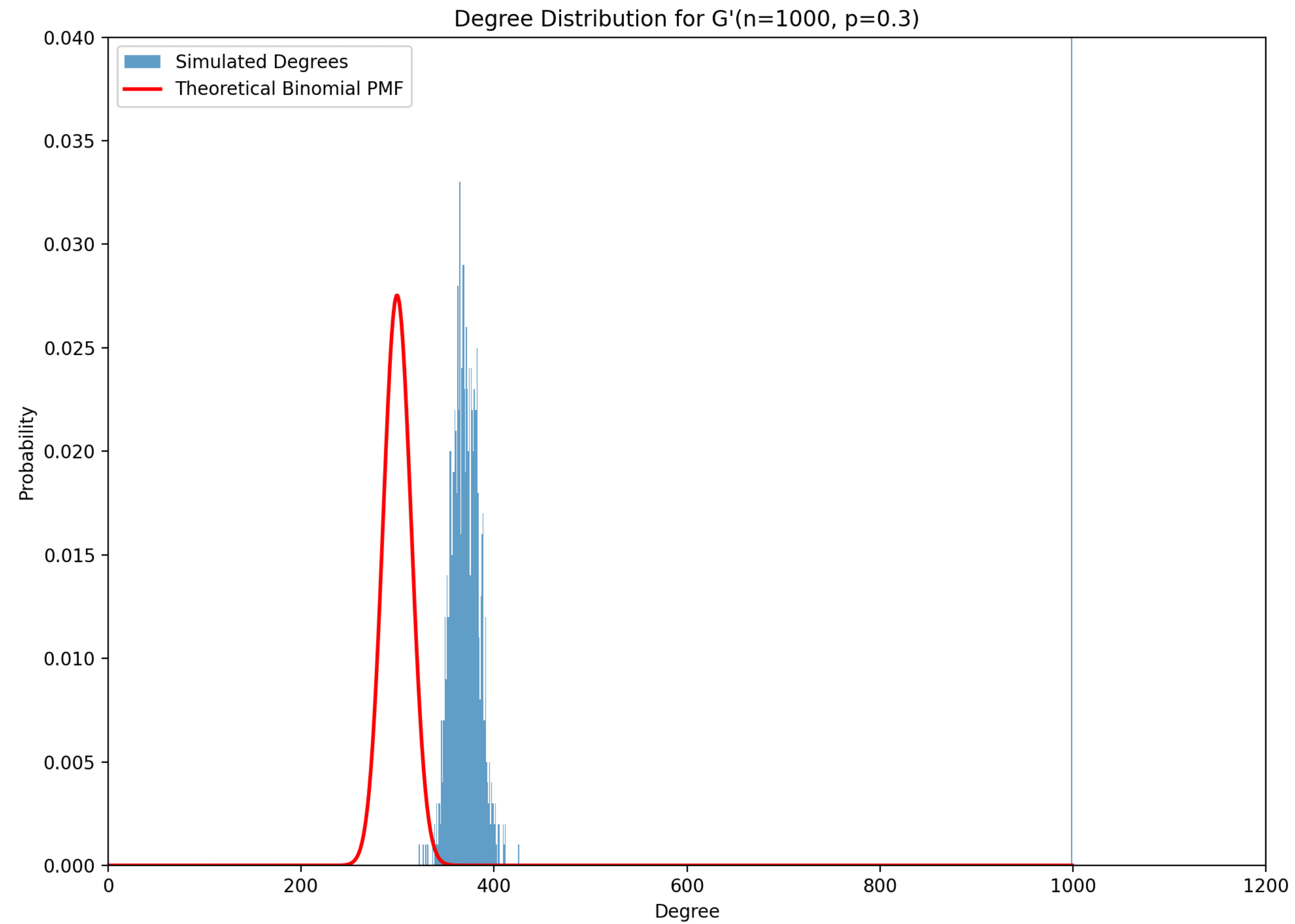


# *Errors and Time Complexity*

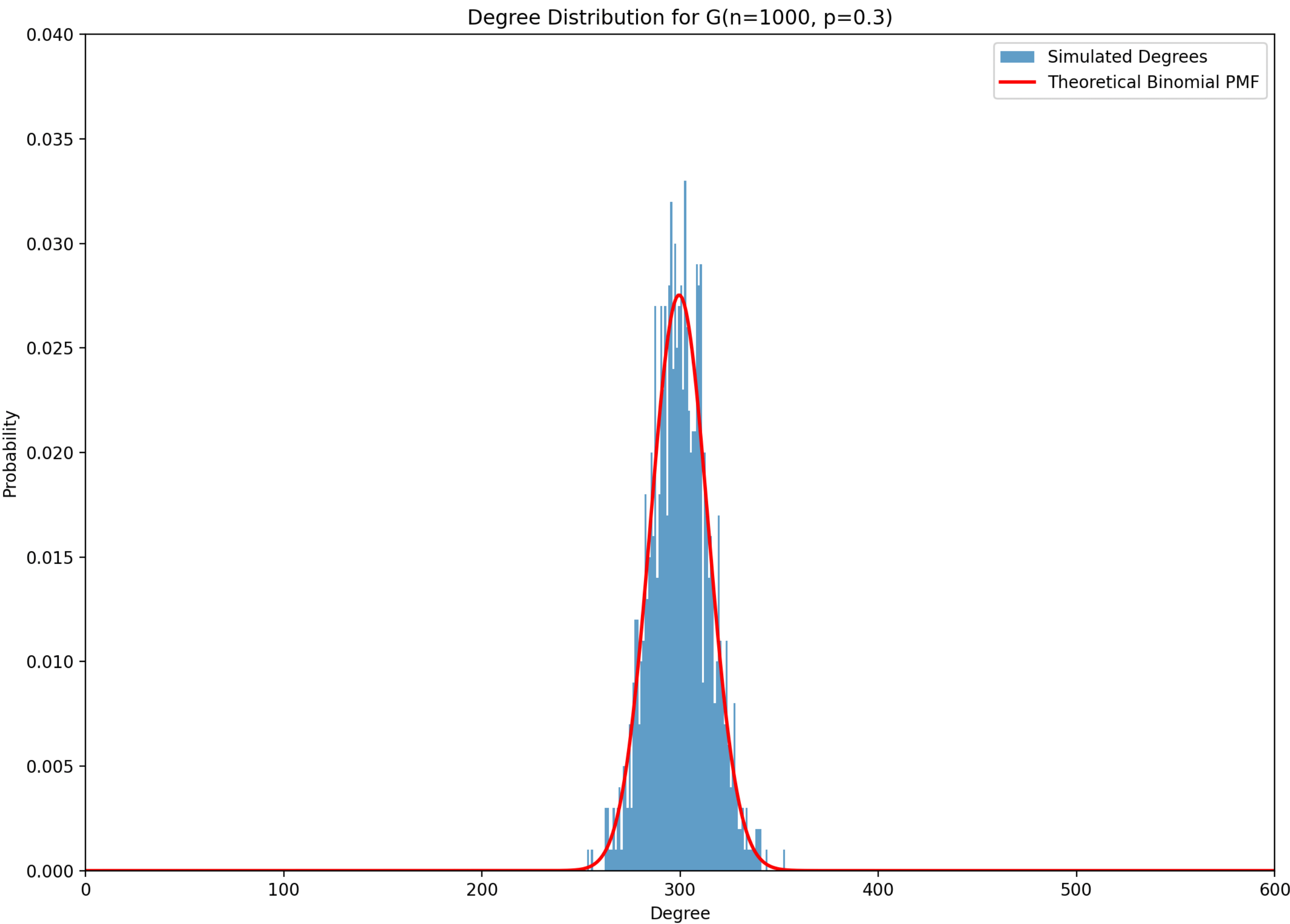
- Error:  $O(\varepsilon)$
- Time complexity:  $O(n)$

Prune Then Mean / Median

# *Degrees of Perturbed Graph*



# Degrees of Unperturbed Graph



# *Errors and Time Complexity*

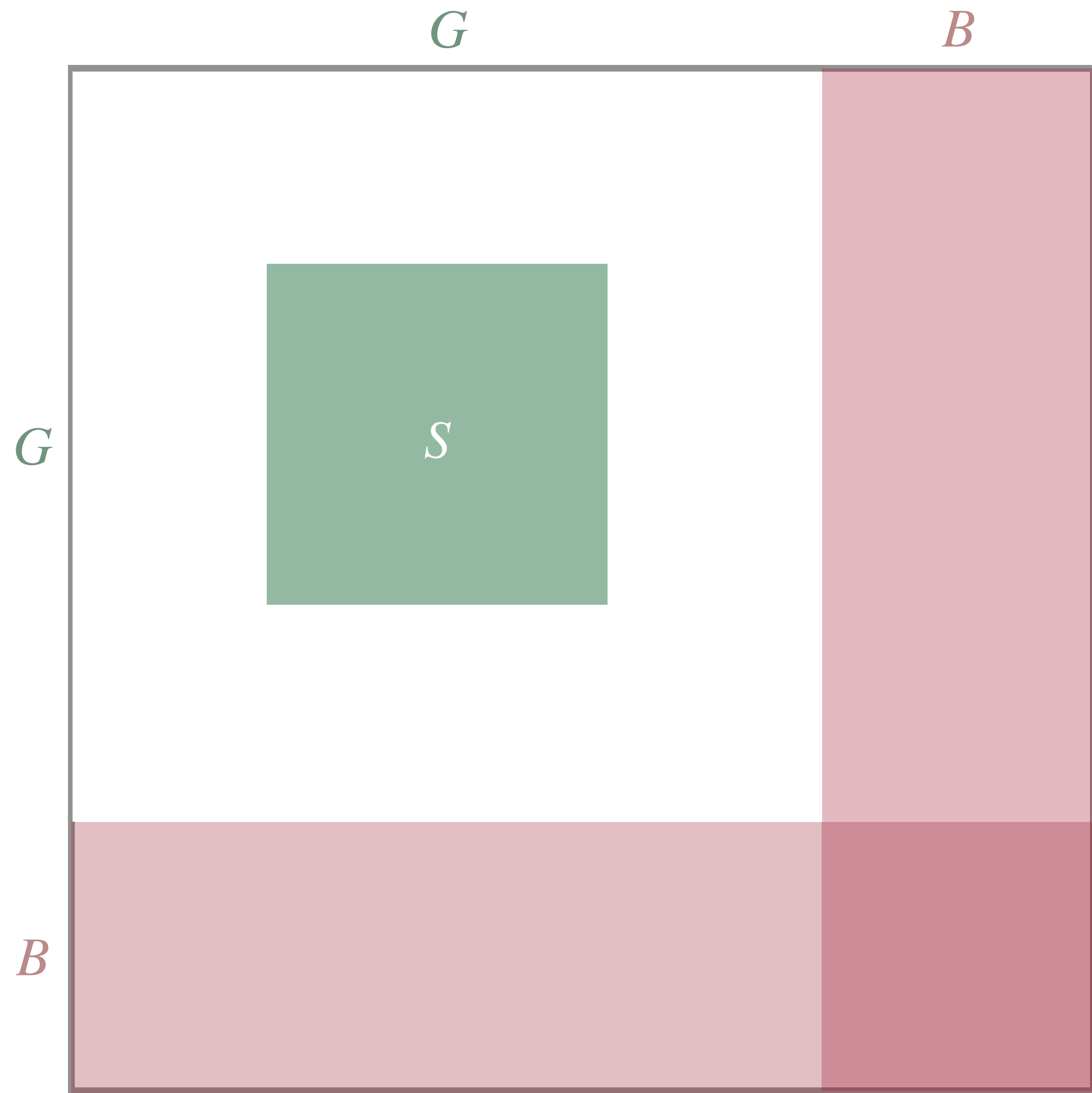
- Error:  $O(\varepsilon^2)$ ,  $O(\varepsilon)$
- Time complexity:  $O(n \log(n\varepsilon))$

# Spectral Method



# Main idea

- We know that an uncorrupted subgraph should have an estimated  $\boldsymbol{p}$  close to the frequency of 1s in the adjacency matrix
- $\|(A - P_S)_{S \times S}\|_2$  is small for uncorrupted subgraph  $S$



# *Spectral Method*

- We start with the entire graph
- Compute our metric  $\|(A - P_S)_{s \times s}\|_2$
- Use the largest eigenvector to find the troublemaker node
- Remove the node
- Repeat

# Our Algorithms

# Notation

- $G = (V, E)$  is our original graph
- $G' = (V, E')$  is our perturbed graph
- $D, D'$  are degrees of  $V, V'$
- $\bar{d}, \bar{d}'$  is the mean of  $D$  and  $D'$
- $\tilde{d}, \tilde{d}'$  is the median of  $D$  and  $D'$

Median - Mean

# Main idea

- Median and mean have errors  $O(\varepsilon)$
- Observed median is ~2x closer to the true  $p$  than mean
- We can exploit this by taking the median minus the mean to find how far away the median is from  $p$

How well can we estimate  $p$ ?



# Proof Sketch

- $\bar{d}' \approx d + \varepsilon(q - p)(2n - 1) - \varepsilon^2 n(q - p)$

- $\tilde{d}' \approx d + \varepsilon n(q - p) - \varepsilon \sqrt{np(1 - p)}$

- Claim:  $\hat{d} := \tilde{d}' - \frac{\bar{d}' - \tilde{d}'}{1 - \varepsilon}$  is good

- $\tilde{d}' - \frac{\bar{d}' - \tilde{d}'}{1 - \varepsilon} = \tilde{d}' - \left( \sum_{k=0}^{\infty} \varepsilon^k \right) (\bar{d}' - \tilde{d}') = d + \frac{1}{1 - \varepsilon} (q - p + \varepsilon \sqrt{np(1 - p)})$

- Then  $\hat{p} = \frac{\hat{d}}{n} = p + \frac{\frac{\varepsilon}{1 - \varepsilon} (q - p + \varepsilon \sqrt{np(1 - p)}) - p}{n} = p + O \left( \frac{\varepsilon p}{(1 - \varepsilon)n} + \frac{\varepsilon^2 \sqrt{p(1 - p)}}{(1 - \varepsilon)\sqrt{n}} \right)$

---

**Algorithm 1** Median - Mean Algorithm

---

**Require:** adjacency matrix  $A$ , epsilon  $\epsilon$

$D \leftarrow$  Degrees of nodes in  $A$

$\bar{p} \leftarrow$  normalized mean

$\tilde{p} \leftarrow$  normalized median

$y \leftarrow \frac{1}{1-\epsilon} |\bar{p} - \tilde{p}|$

**if**  $\bar{p} \leq \tilde{p}$  **then**

**return**  $\tilde{p} - y$

**else**

**return**  $\bar{p} - y$

**end if**

---

# *Errors and Time Complexity*

- Error:  $O\left(\frac{\varepsilon p}{(1 - \varepsilon)n} + \frac{\varepsilon^2 \sqrt{p(1 - p)}}{(1 - \varepsilon)\sqrt{n}}\right)$
- Time complexity:  $O(n)$

# Variance Method

# Variance Method

- The variance of degrees on an unperturbed graph should be:  $(n - 1)p(1 - p)$
- Variance significantly increases on a perturbed graph with an adversary

# Variance Algorithm

- Remove the  $\epsilon n$  bad nodes
- Find the node that decreases the variance the most
- Remove that node
- Repeat

---

**Algorithm 2** Variance Algorithm

---

**Require:** adjacency matrix  $A$ , epsilon  $\epsilon$

$n \leftarrow$  number of nodes

**while**  $t = 0, t < \epsilon n$  **do**

**for** node  $x$  in  $A$  **do**

        Modify degrees of nodes after removing  $x$

        Compute variance on subgraph degrees

**end for**

$v \leftarrow$  node with largest variance deviation

$A \leftarrow A \setminus v$

**end while**

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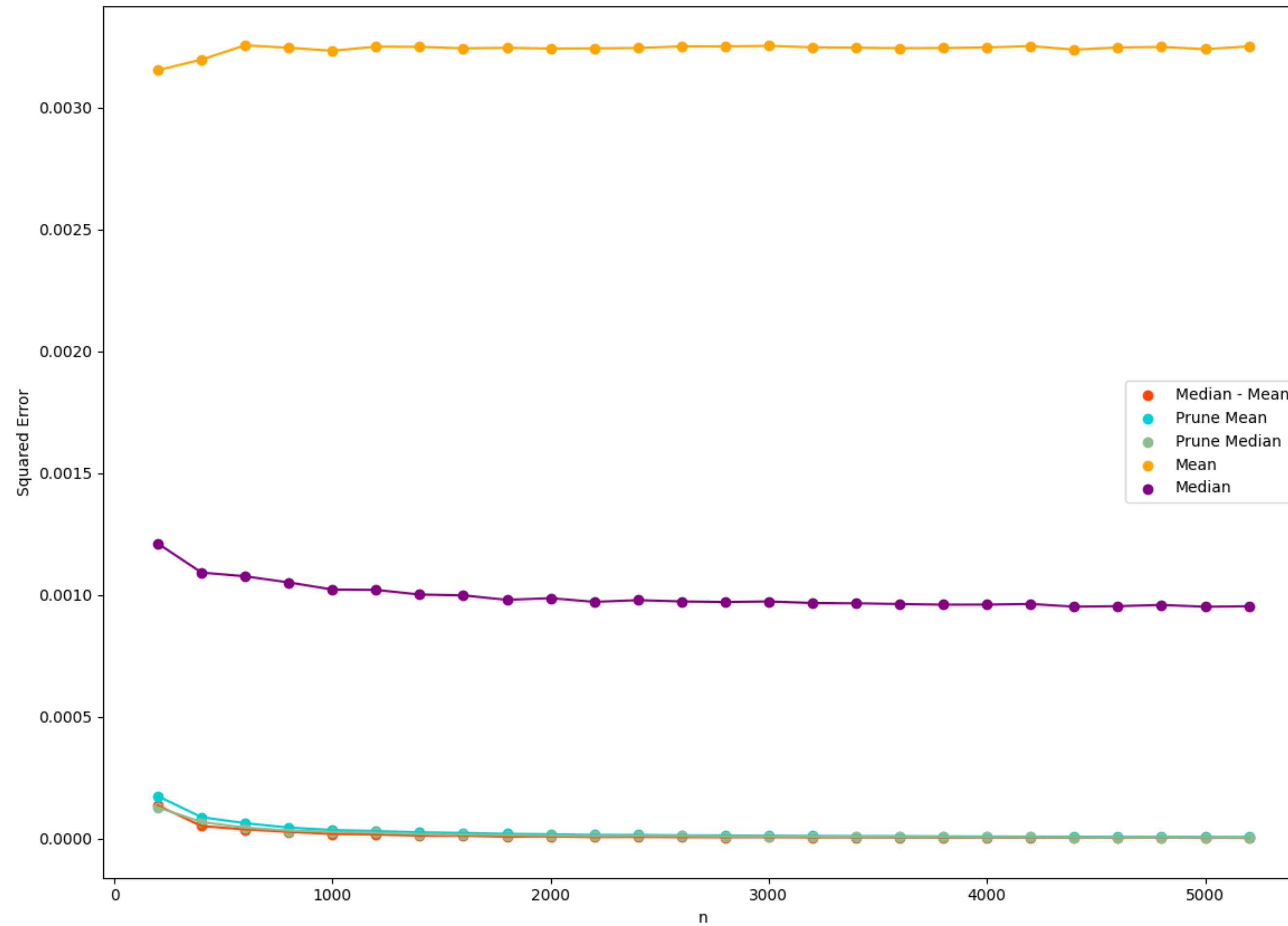
# *Errors and Time Complexity*

- Error: ?
- Time complexity:  $O(\epsilon n^3)$

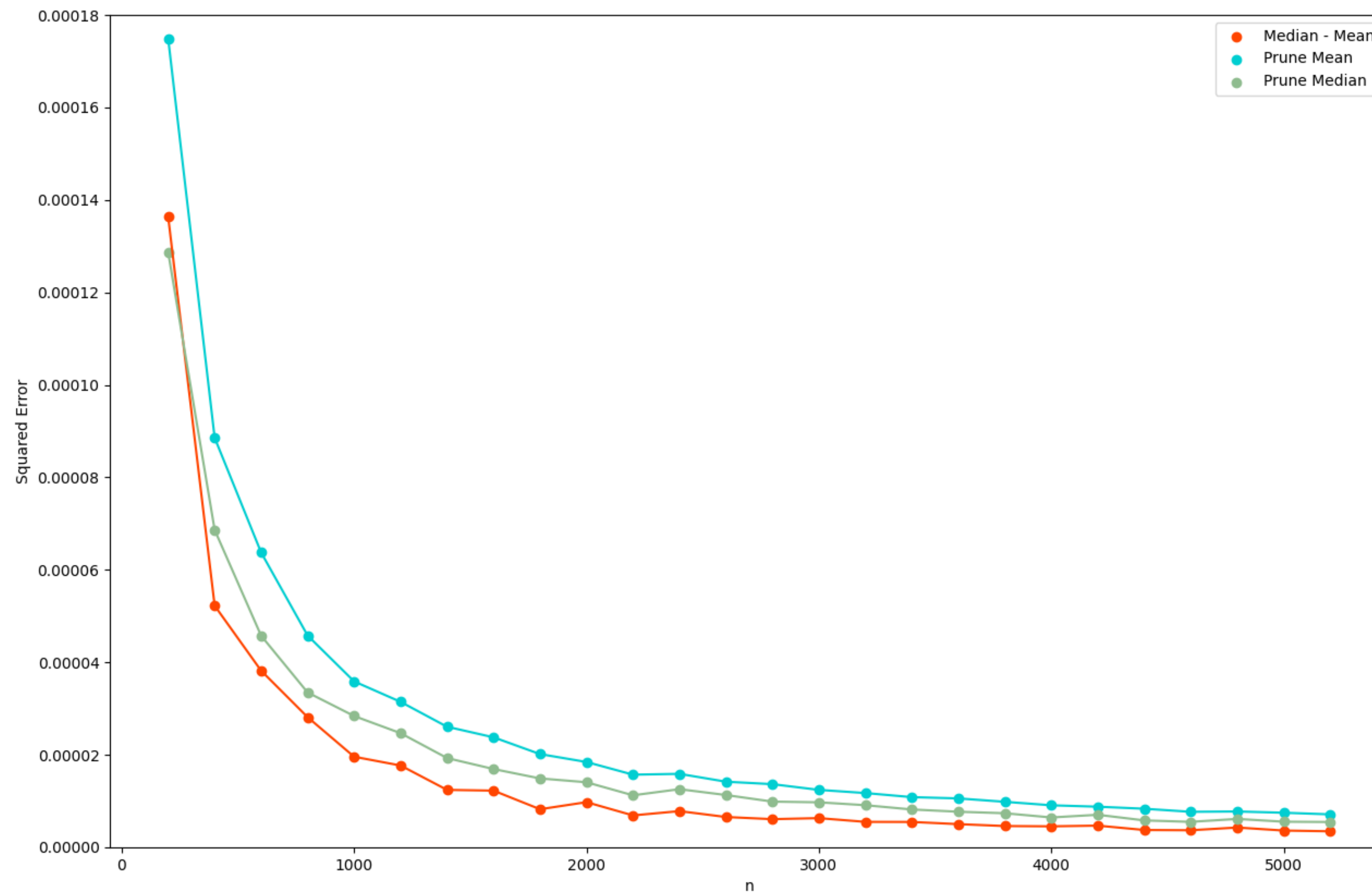


# Results

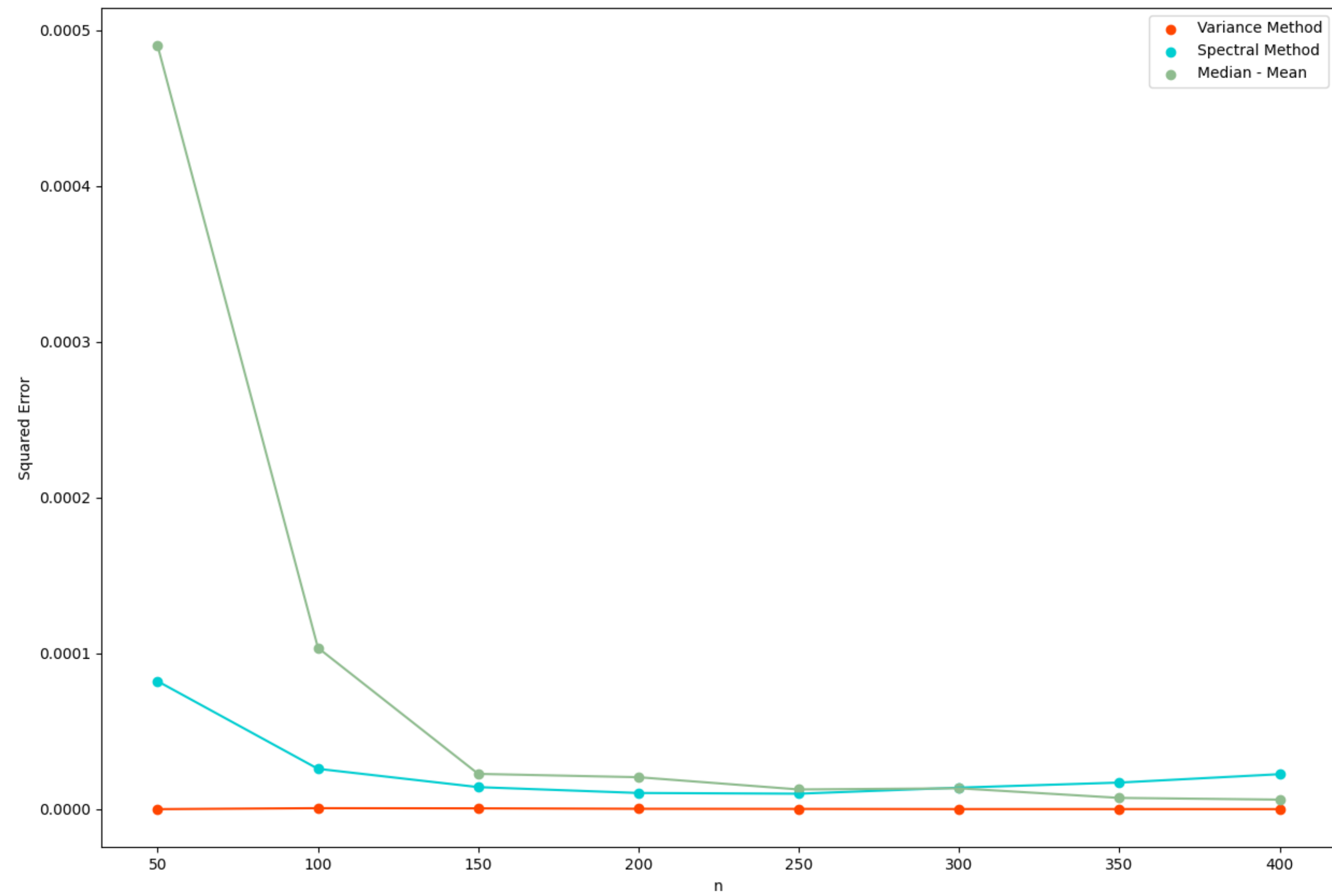
# Results (Median - Mean)



# Results (Median - Mean)



# Results (Variance)



# Results (Variance)



# Results

Methods	Error	Runtime	Authors
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Median – Mean	$O\left(\frac{\varepsilon p}{(1-\varepsilon)n} + \frac{\varepsilon^2\sqrt{p(1-p)}}{(1-\varepsilon)\sqrt{n}}\right)^*$	$O(n)$	Us
Variance Method	?	$O(\varepsilon n^3)$	Us

# Next Steps

- Update error bounds on Median-Mean algorithm for general adversary
- Attempt to prove error bounds for variance method

Thank you!