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Inter-University Research Institute Corporation /
Research Organization of Information and Systems
National Institute of Informatics



Significant Pattern Mining on Graphs

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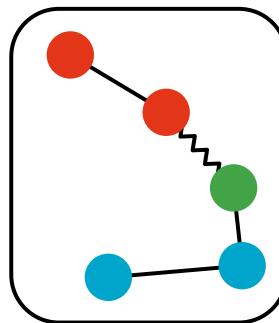
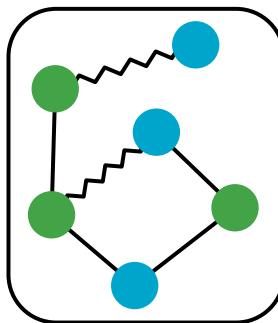
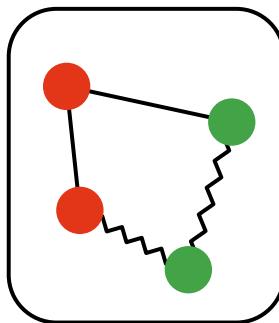
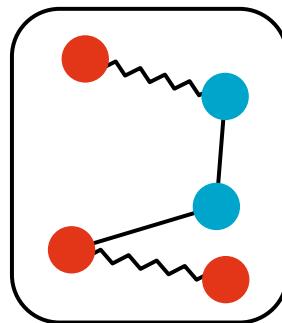
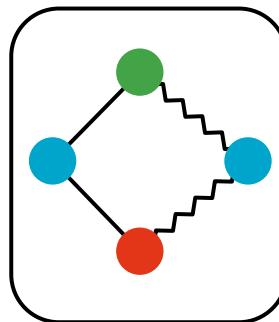
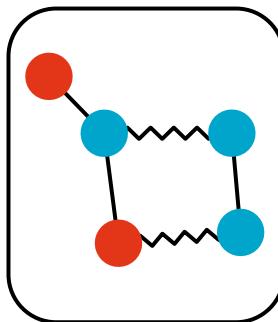
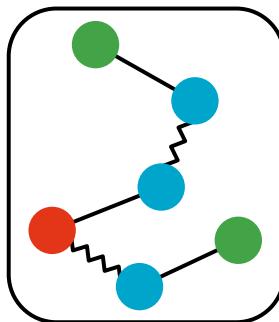
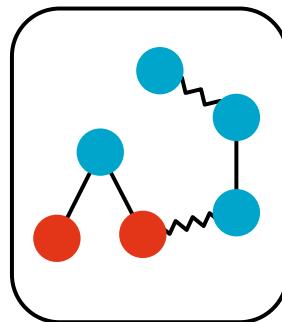
Literature

- Sugiyama, M., Llinares-López, F., Kasenburg, N., Borgwardt, K.:
Significant Subgraph Mining with Multiple Testing Correction,
SIAM SDM 2015
- Llinares-López, F., Sugiyama, M., Papaxanthos, L., Borgwardt, K.:
Fast and Memory-Efficient Significant Pattern Mining via Permutation Testing,
ACM SIGKDD 2015

Subgraph Mining

- Find interesting **subgraphs** from graph databases

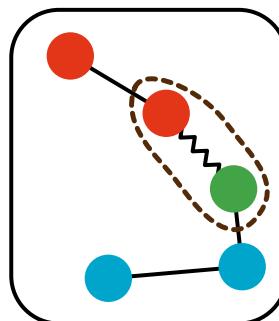
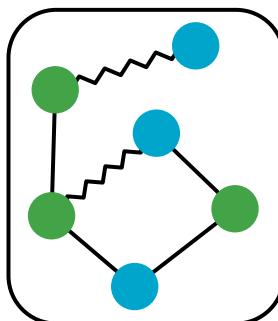
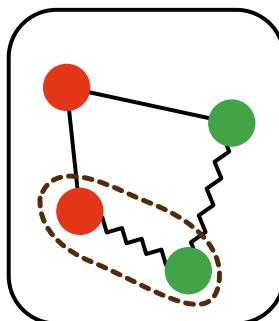
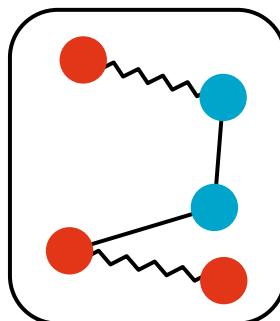
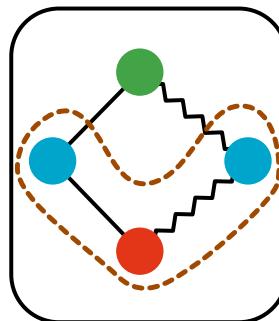
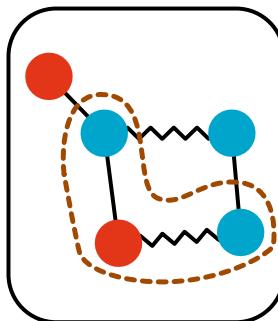
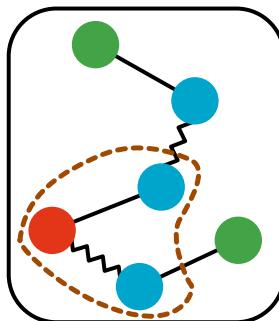
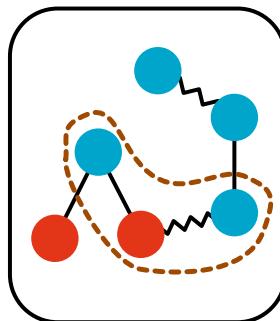
Database



Subgraph Mining

- Find interesting **subgraphs** from graph databases

Database



Subgraph



Support: 4

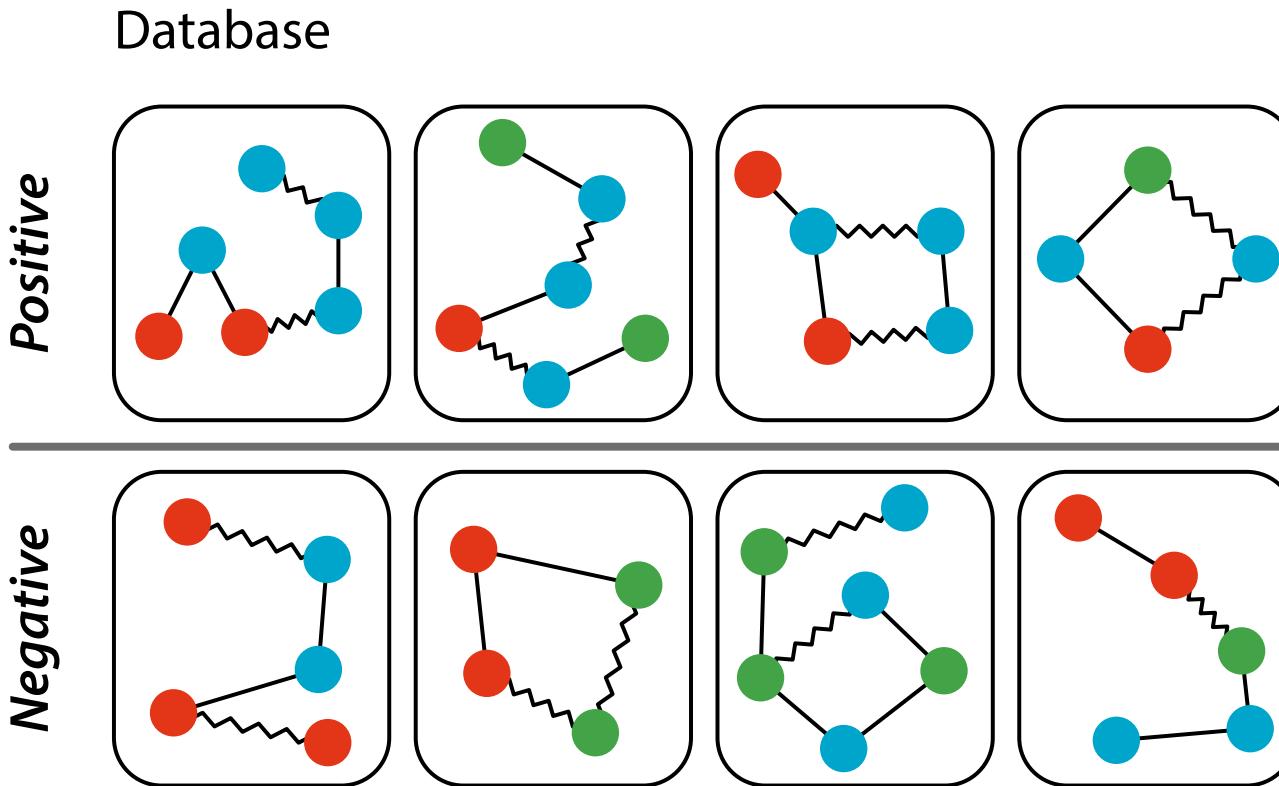


Support: 2



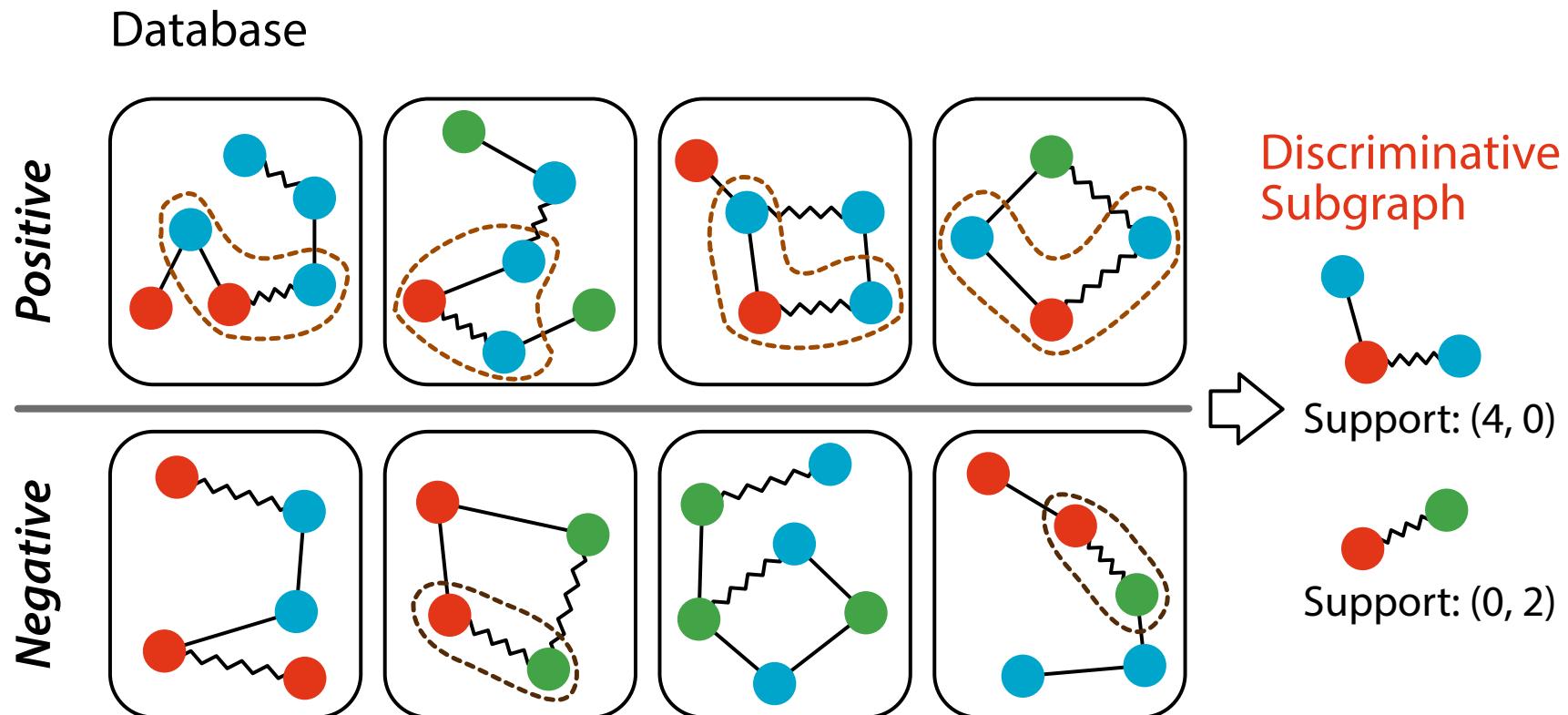
Discriminative Subgraph Mining

- Find **discriminative subgraphs** from **supervised** data
(e.g. Drug discovery)



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Challenges and Solutions

- In discriminative subgraph mining:
 1. How to measure the **discriminability** of subgraphs?
 2. How to enumerate all discriminative subgraphs?

Challenges and Solutions

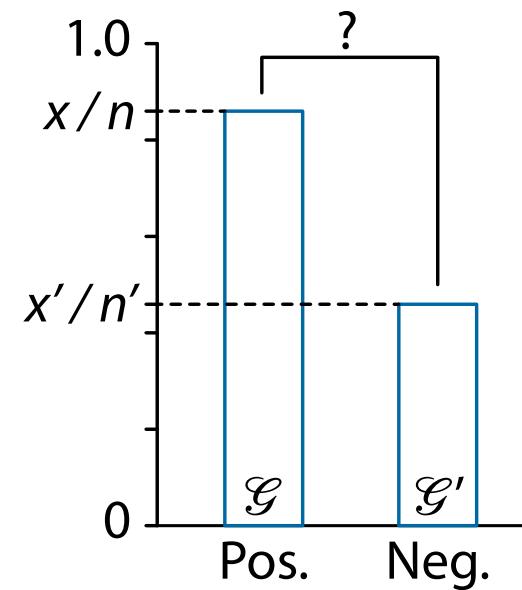
- In discriminative subgraph mining:
 1. How to measure the **discriminability** of subgraphs?
 2. How to enumerate all discriminative subgraphs?
- *Answer to 1:*
 - Compute the **p-value** via **statistical hypothesis testing**
 - Discriminative subgraph \iff (Statistically) Significant subgraph
- *Answer to 2:*
 - Integrate evaluation of discriminability and enumeration of subgraphs

Computing p -value of Subgraph

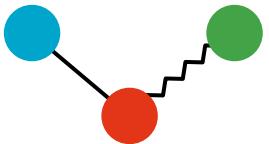
- Given positive and negative sets of graphs $\mathcal{G}, \mathcal{G}'$
 - $|\mathcal{G}| = n, |\mathcal{G}'| = n' (n \leq n')$
- The *p-value* of each subgraph H is determined by the *Fisher's exact test*
 - $x = |\{G \in \mathcal{G} \mid H \subseteq G\}|$

	Occ.	Non-occ.	Total
\mathcal{G} (Pos.)	x	$n - x$	n
\mathcal{G}' (Neg.)	x'	$n' - x'$	n'
Total	$x + x' = \sigma$	$(n - x) + (n' - x')$	$n + n'$

Support



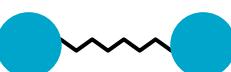
Multiple Testing



	Occ.	Non-occ.	Total
Positive	4	0	4
Negative	2	2	4
Total	6	2	8

Fisher's exact test: $p\text{-value} = 0.429$

Multiple Testing



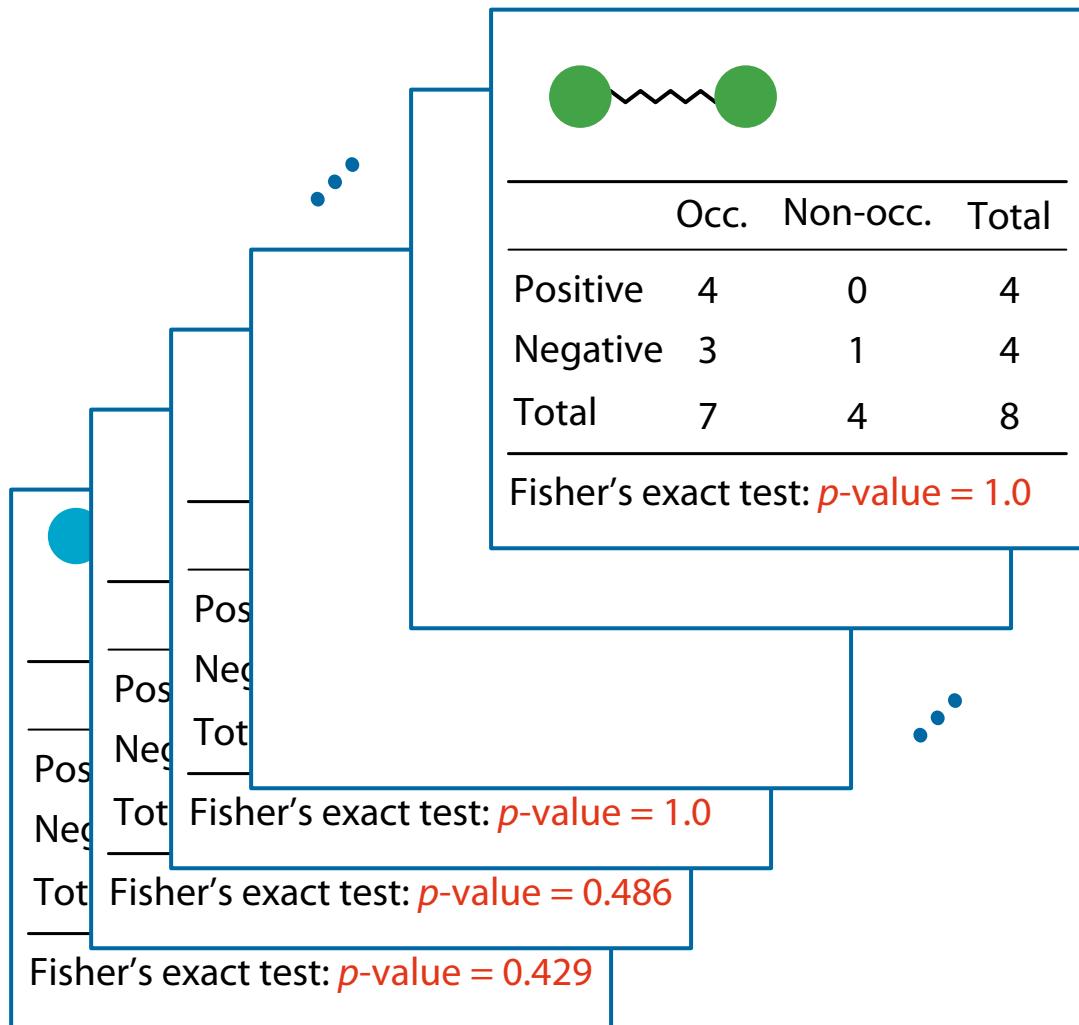
	Occ.	Non-occ.	Total
Positive	3	1	4
Negative	1	3	4
Total	4	4	8
Tot	Fisher's exact test: <i>p</i> -value = 0.486		
	Fisher's exact test: <i>p</i> -value = 0.429		

Multiple Testing

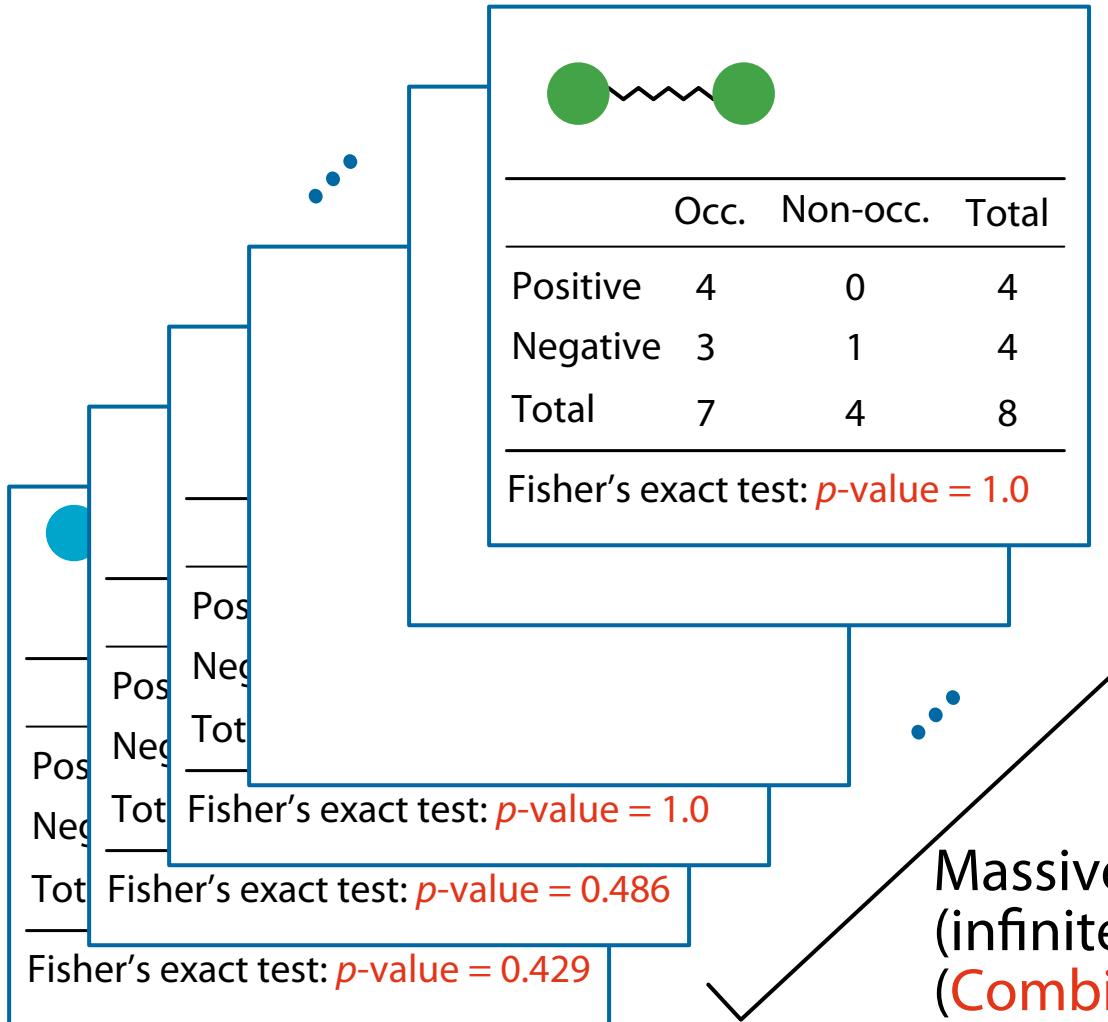


	Occ.	Non-occ.	Total
Positive	2	2	4
Negative	2	2	4
Total	4	4	8
Fisher's exact test: <i>p</i> -value = 1.0			
Fisher's exact test: <i>p</i> -value = 0.486			
Fisher's exact test: <i>p</i> -value = 0.429			

Multiple Testing



Multiple Testing



Task: Enumerate all significant subgraphs while controlling the FWER

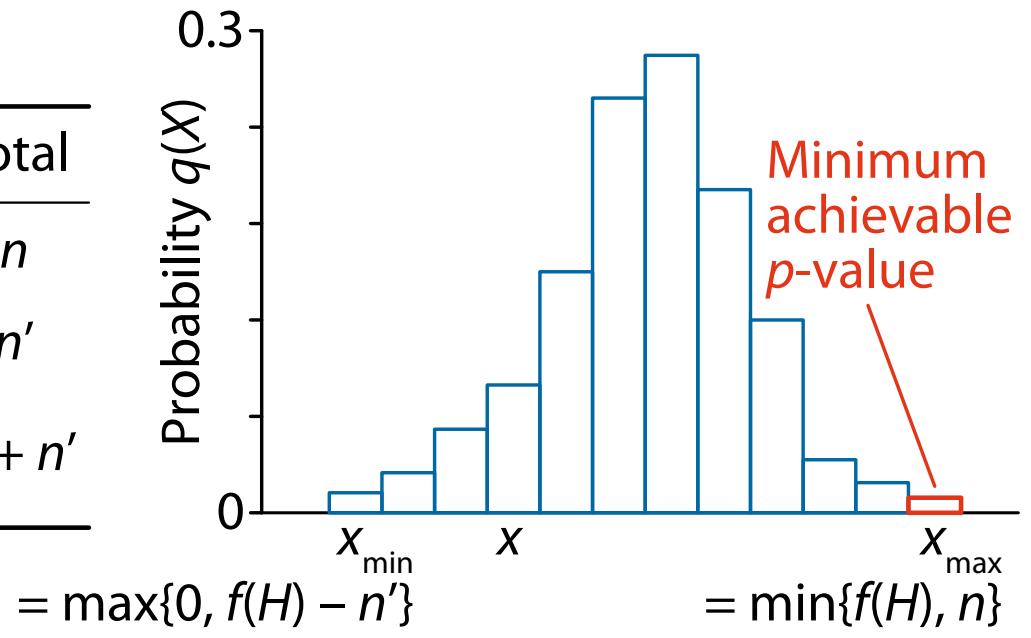
Massive number of (infinitely many) subgraphs (Combinatorial explosion!)

Minimum Achievable p -value $\Psi(\sigma)$

- Consider the **minimum achievable p -value $\Psi(\sigma)$** of a subgraph H for its **support** $\sigma = |\{X \in \mathcal{X} \cup \mathcal{X}' \mid H \subseteq X\}|$
 - $\Psi(\sigma) = \min\{p(x) \mid x_{\min} \leq x \leq x_{\max}\}$
 - $x_{\min} = \max\{0, \sigma - n'\}$, $x_{\max} = \min\{\sigma, n\}$

	Occ.	Non-occ.	Total
\mathcal{G} (Pos.)	x	$n - x$	n
\mathcal{G}' (Neg.)	x'	$n' - x'$	n'
Total	$x + x' = \sigma$	$(n - x) + (n' - x')$	$n + n'$

Support



Computing $\Psi(\sigma)$

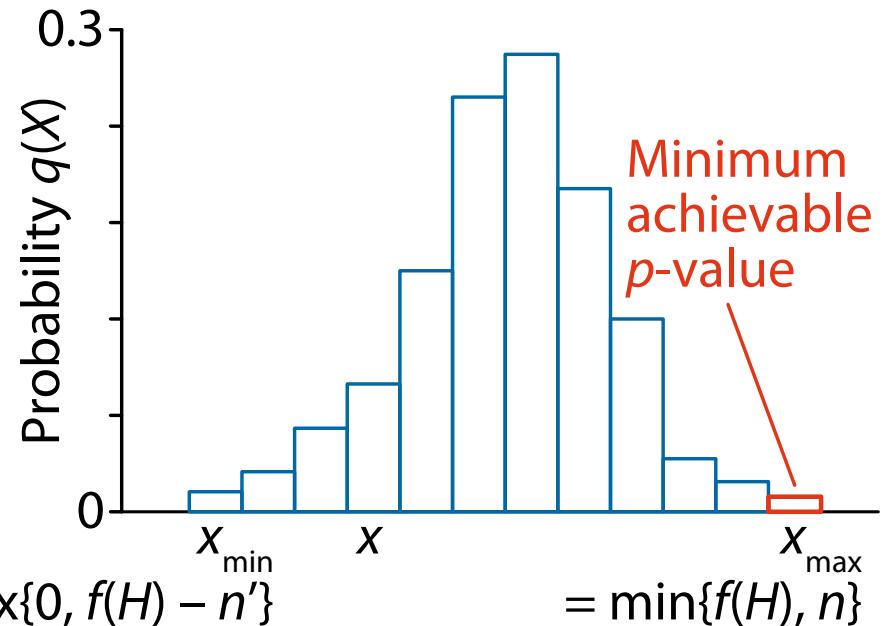
- Consider the **minimum achievable p -value $\Psi(\sigma)$** of a subgraph H for its **support** $\sigma = |\{X \in \mathcal{X} \cup \mathcal{X}' \mid H \subseteq X\}|$

$$\Psi(\sigma) = \binom{n}{\sigma} / \binom{n+n'}{\sigma}$$

	Occ.	Non-occ.	Total
\mathcal{G} (Pos.)	σ	$n - \sigma$	n
\mathcal{G}' (Neg.)	0	n'	n'
Total	σ	$(n - \sigma) + n'$	$n + n'$

Most biased case ($\sigma < n$)

$$= \max\{0, f(H) - n'\}$$



Testability

- Consider the **minimum achievable p -value** $\Psi(\sigma)$ of a subgraph H for its **support** $\sigma = |\{X \in \mathcal{X} \cup \mathcal{X}' \mid H \subseteq X\}|$

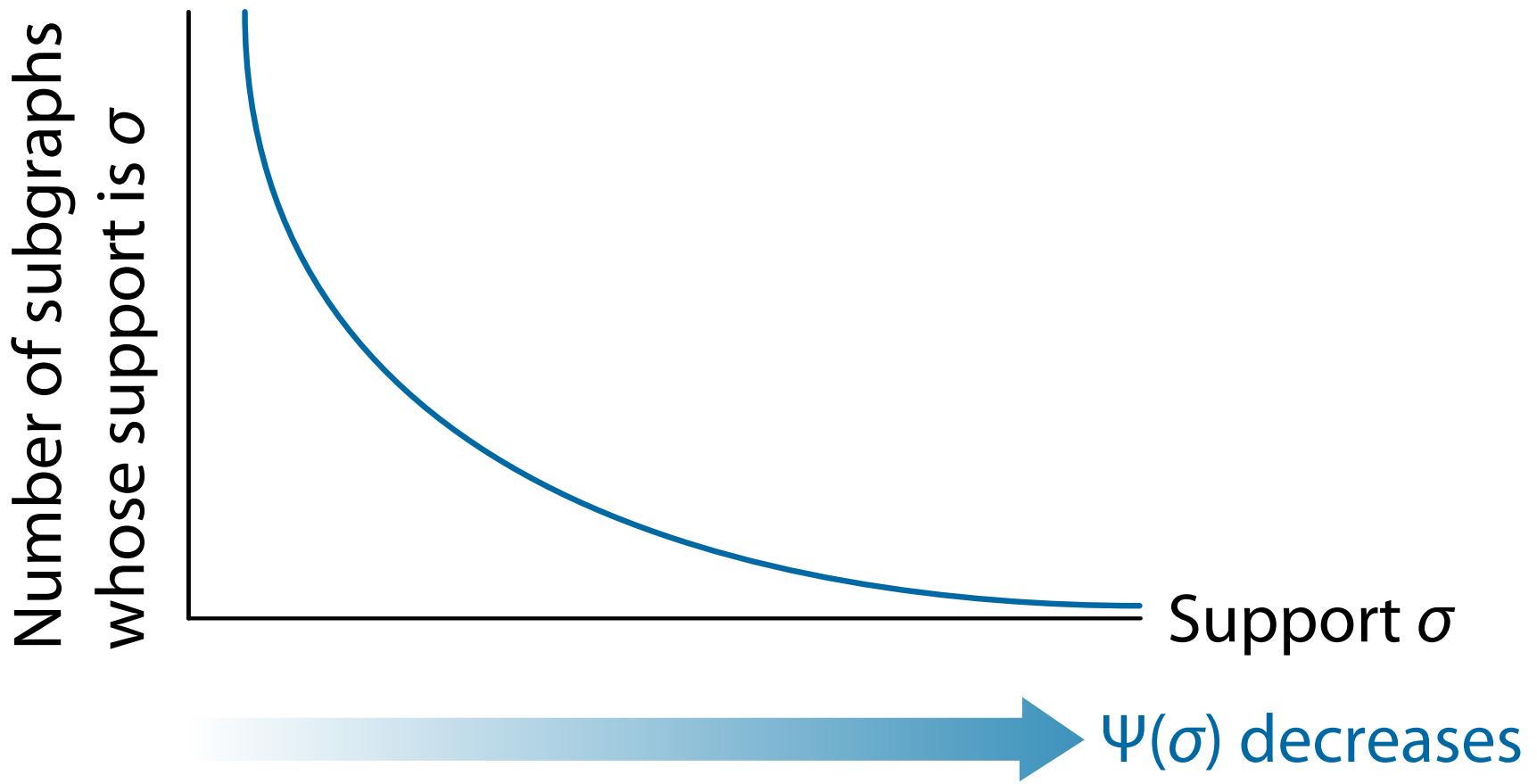
$$\Psi(\sigma) = \binom{n}{\sigma} / \binom{n + n'}{\sigma}$$

- Tarone (1990) pointed out (and Terada et al. (2013) revisited):

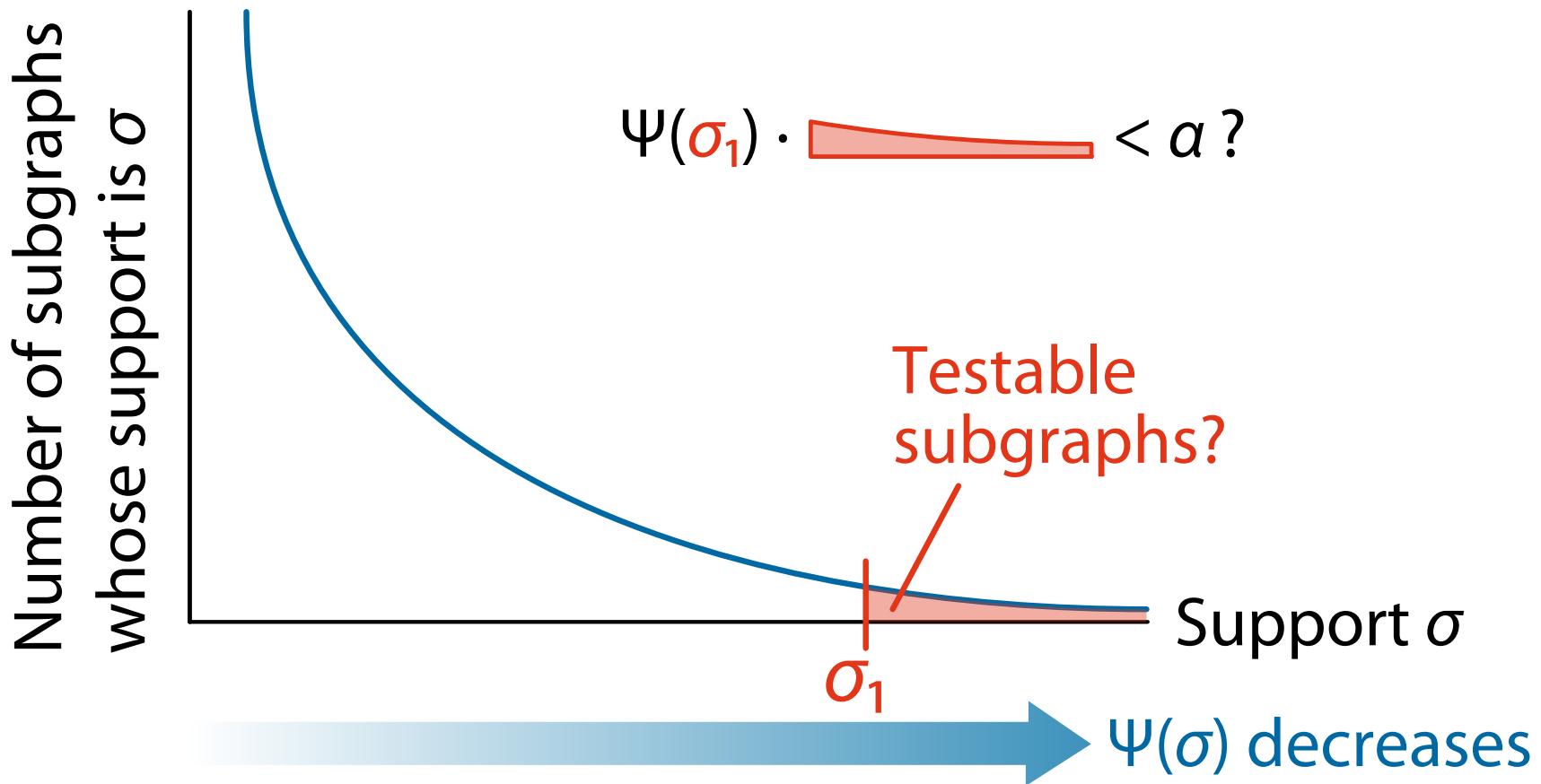
*For a subgraph H with its support σ , if the minimum achievable p -value $\Psi(\sigma)$ is larger than the significance threshold, this is **untestable** and we can ignore it*

- Significance threshold = $\alpha / [\# \text{ testable subgraphs}]$
- Untestable subgraphs can never be significant

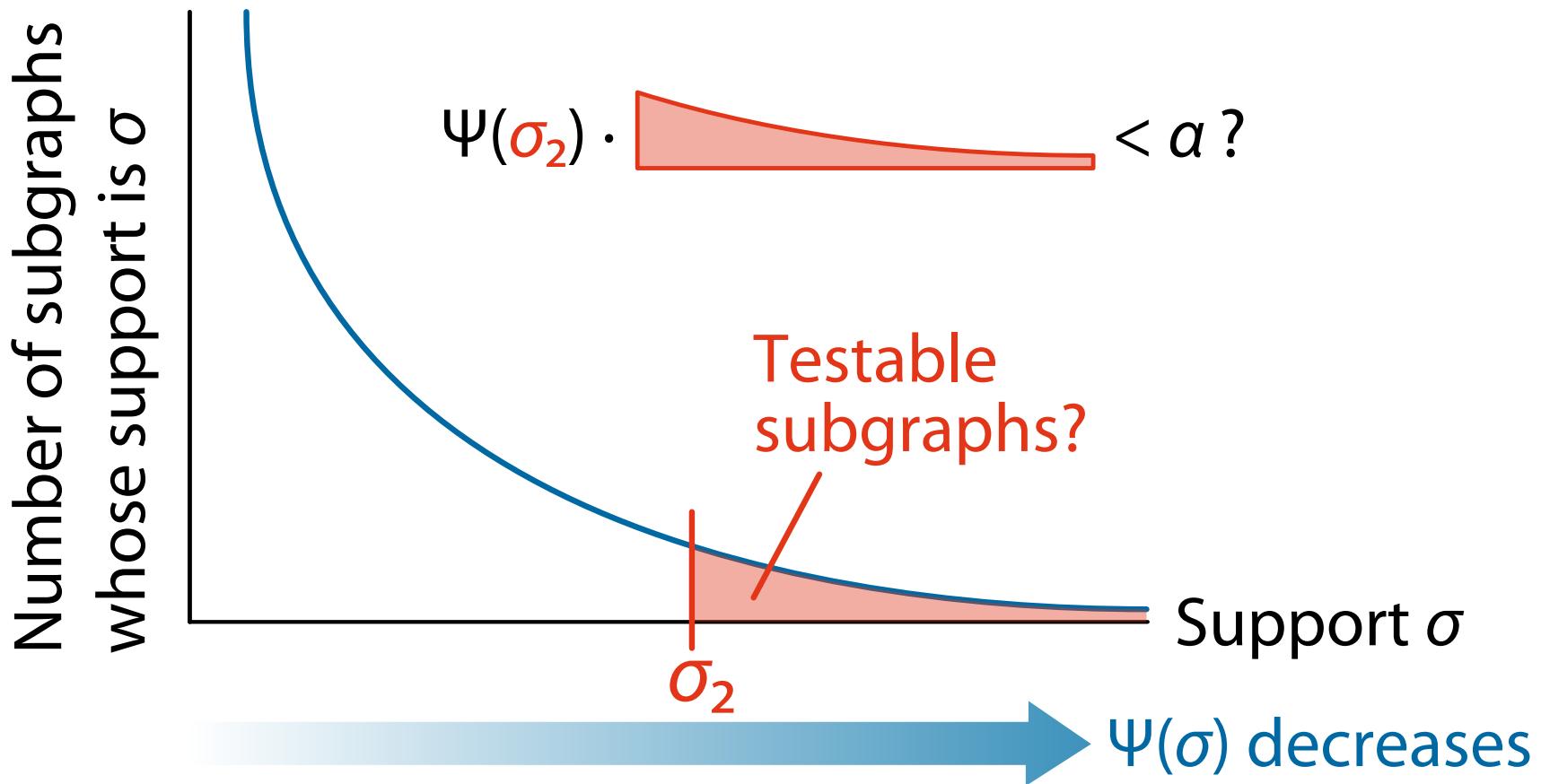
Finding Testable Subgraphs



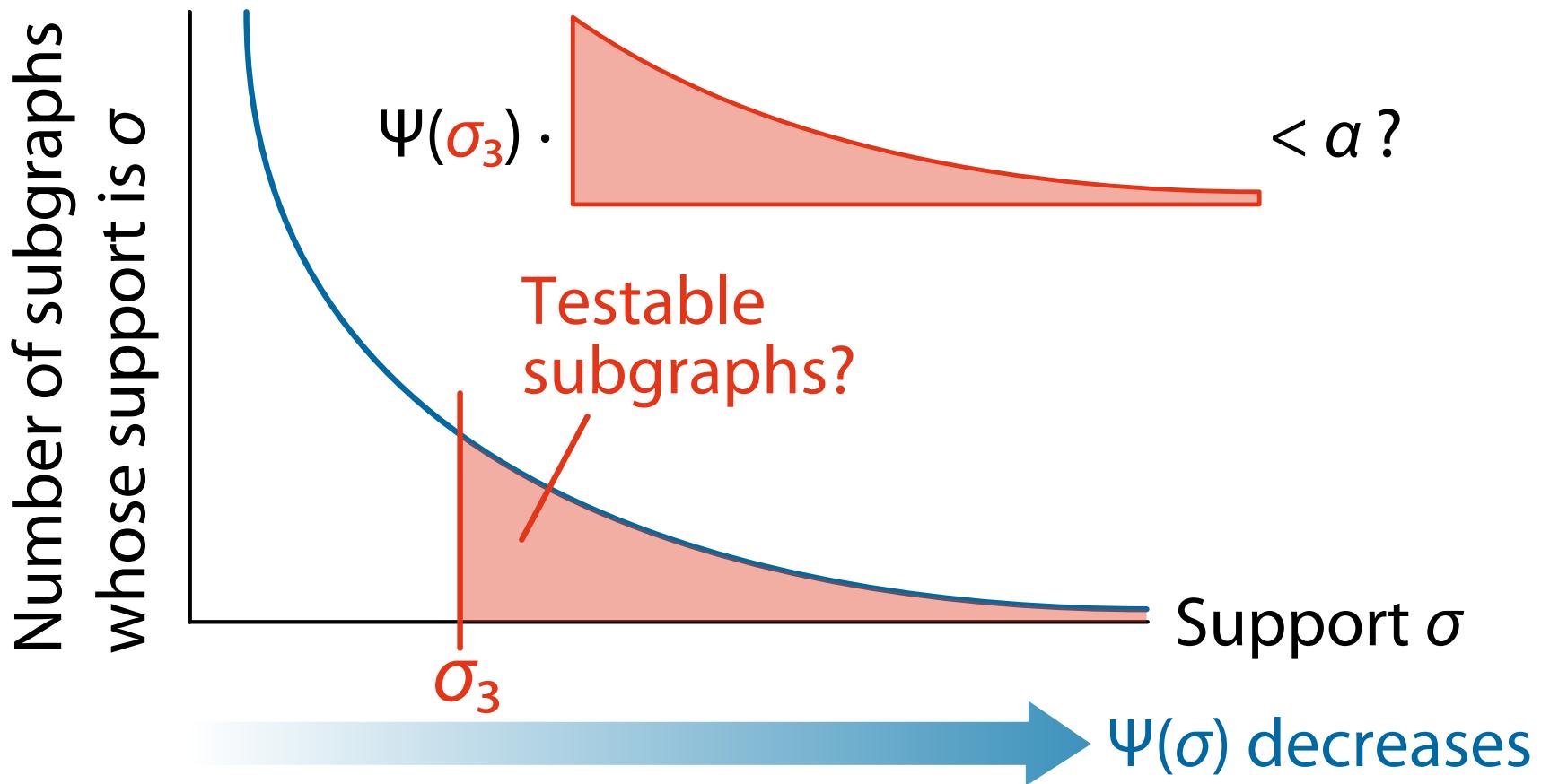
Finding Testable Subgraphs



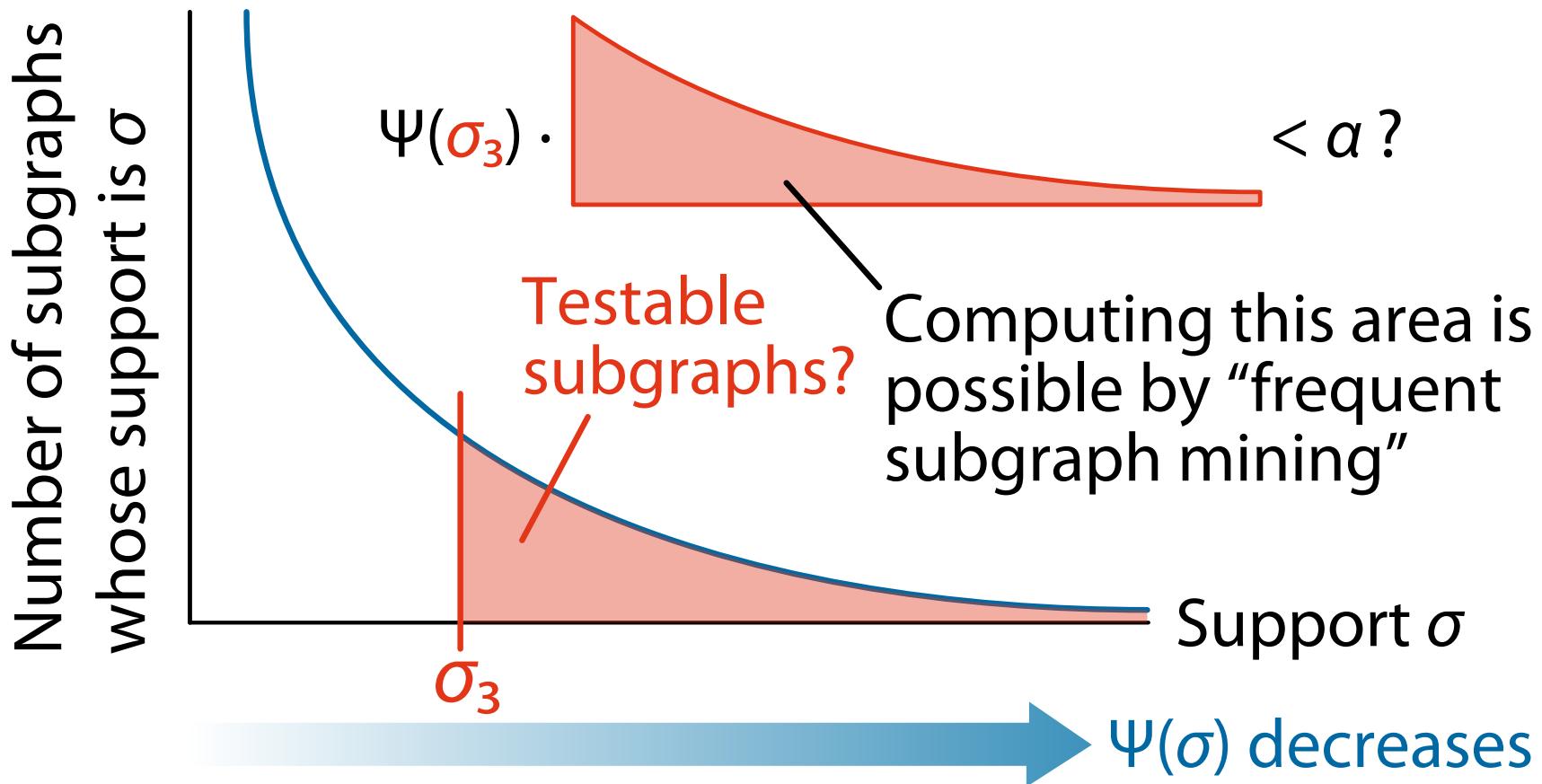
Finding Testable Subgraphs



Finding Testable Subgraphs



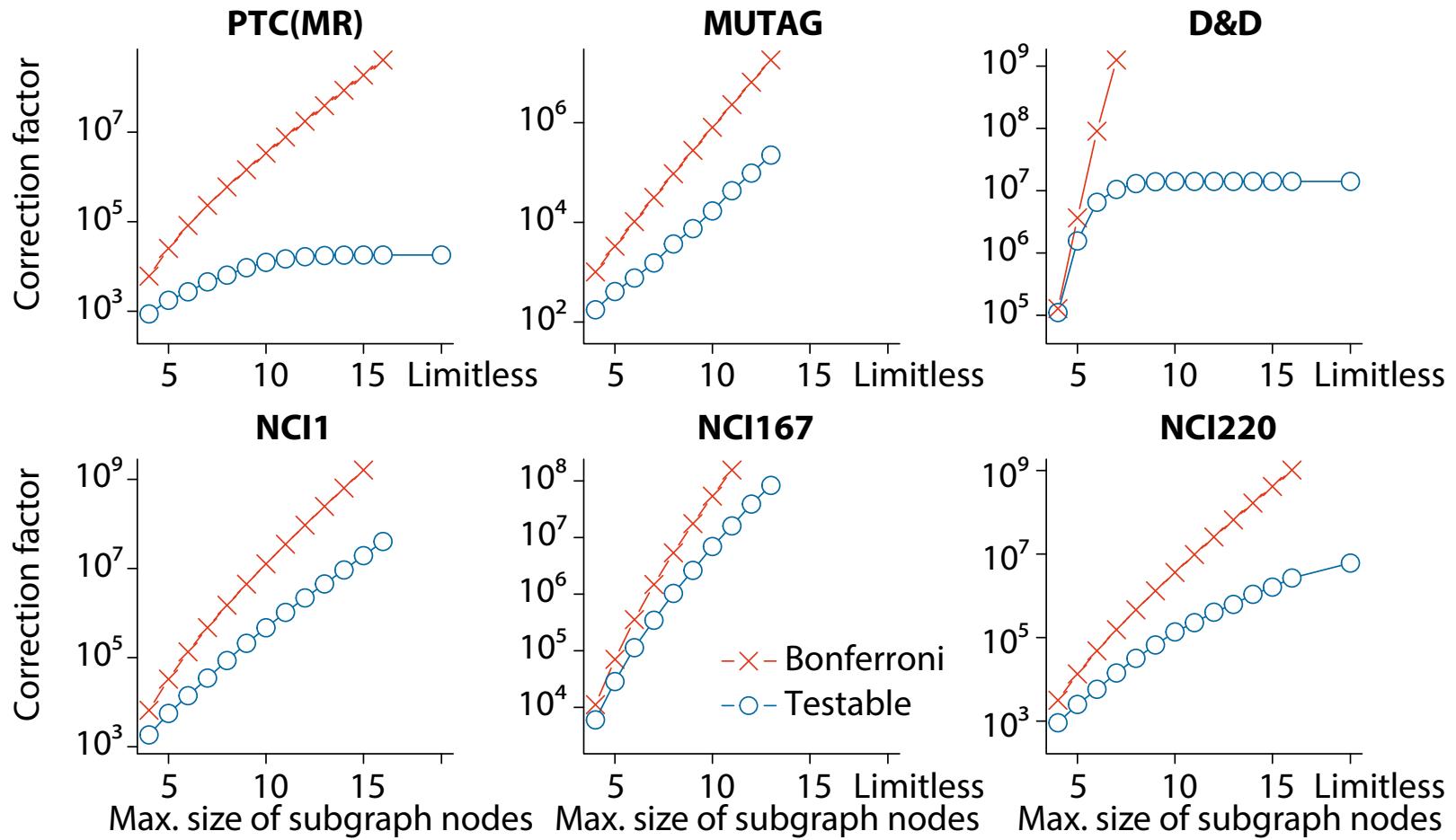
How to Find Testable Subgraphs?



Datasets

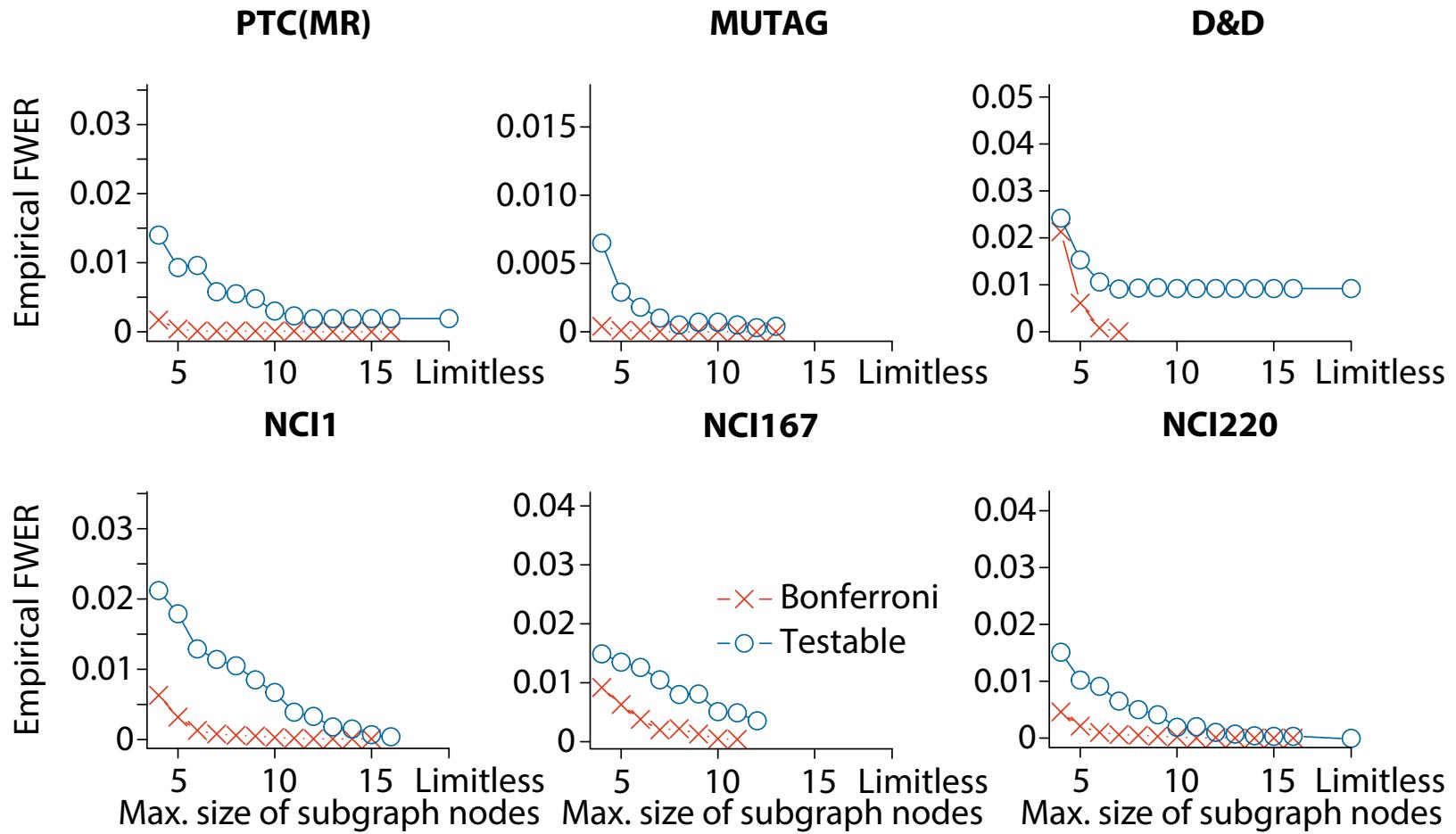
Dataset	Size	#positive	avg. $ V $	avg. $ E $	max $ V $	max $ E $
PTC (MR)	584	181	31.96	32.71	181	181
MUTAG	188	125	17.93	39.59	28	66
D&D	1178	691	284.32	715.66	5748	14267
NCI1	4208	2104	60.12	62.72	462	468
NCI167	80581	9615	39.70	41.05	482	478
NCI220	900	290	46.87	48.52	239	255

Testable Subgraphs



from [Sugiyama et al. SDM2015]

FWER Is Still Too Low!



from [Sugiyama et al. SDM2015]

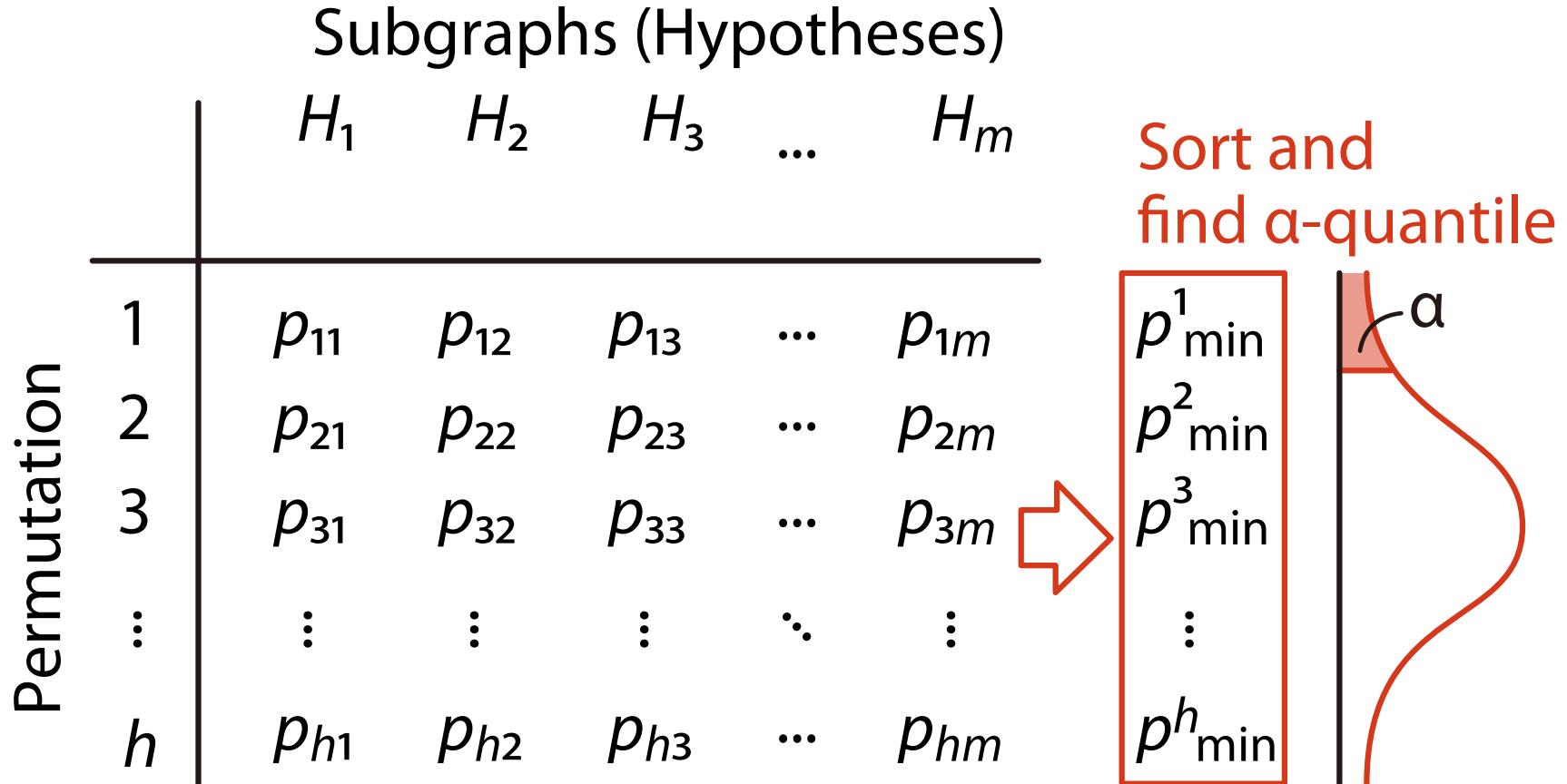
Take Dependencies into Account

- **Problem:** Dependencies between subgraphs are not considered
- **Solution:** Permutation test
 - Repeat random permutation of class labels ($10^3 \sim 10^4$ times)
 - Get the null distribution of p -values
 - The optimal correction factor can be obtained

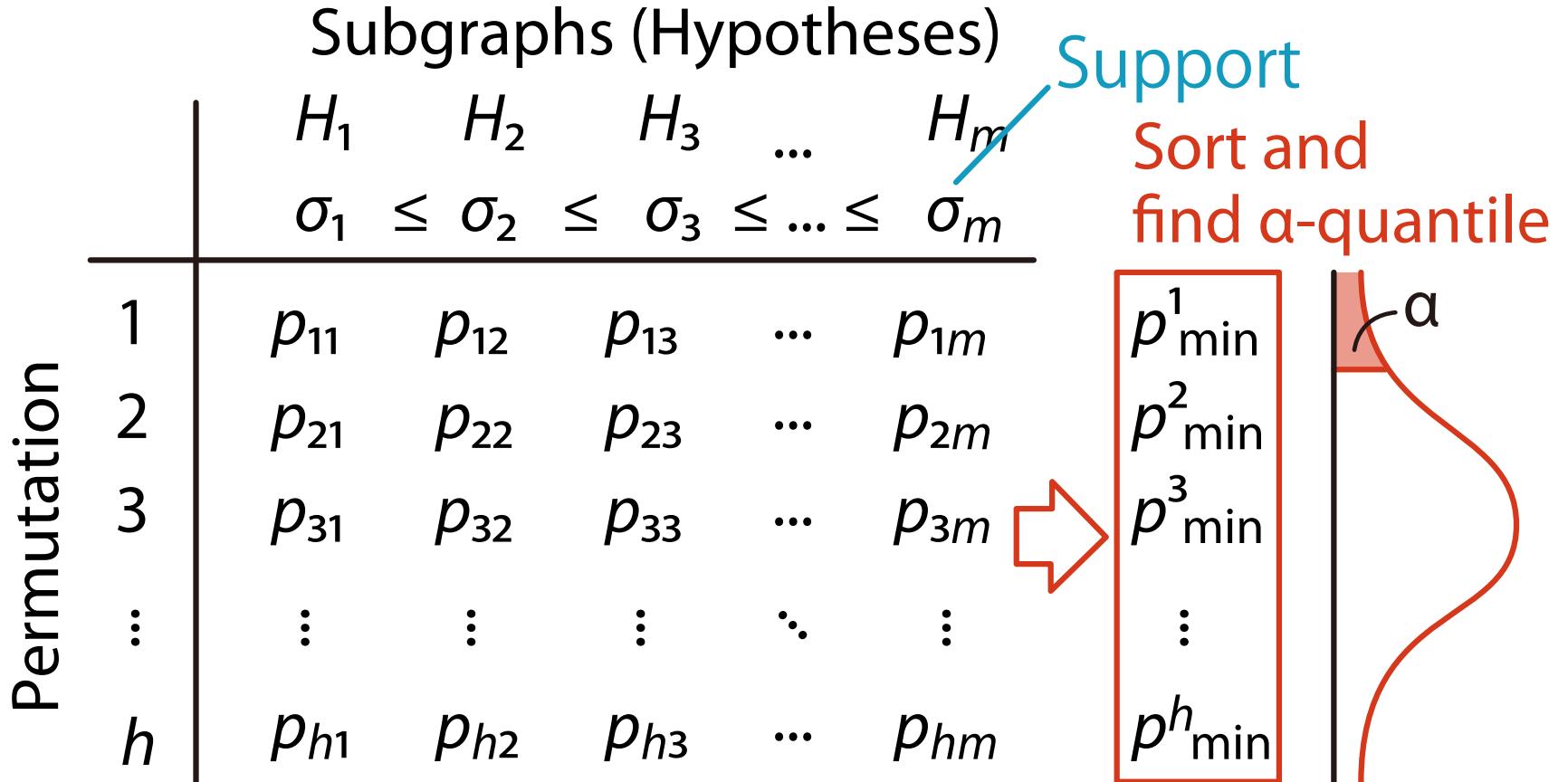
Westfall-Young Permutation

1. Randomly permute class labels
2. Compute p -values for all subgraphs using the permuted class labels
3. Find the minimum p -value p_{\min} among them
 - Number of false positives $> 0 \iff p_{\min} < \delta$
4. Repeat steps 1 to 3 h times and obtain $p_{\min}^1, p_{\min}^2, \dots, p_{\min}^h$
 - $\text{FWER}(\delta) \approx |\{i : p_{\min}^i \leq \delta\}| / h$
5. δ^* is the α -quantile of $p_{\min}^1, p_{\min}^2, \dots, p_{\min}^h$

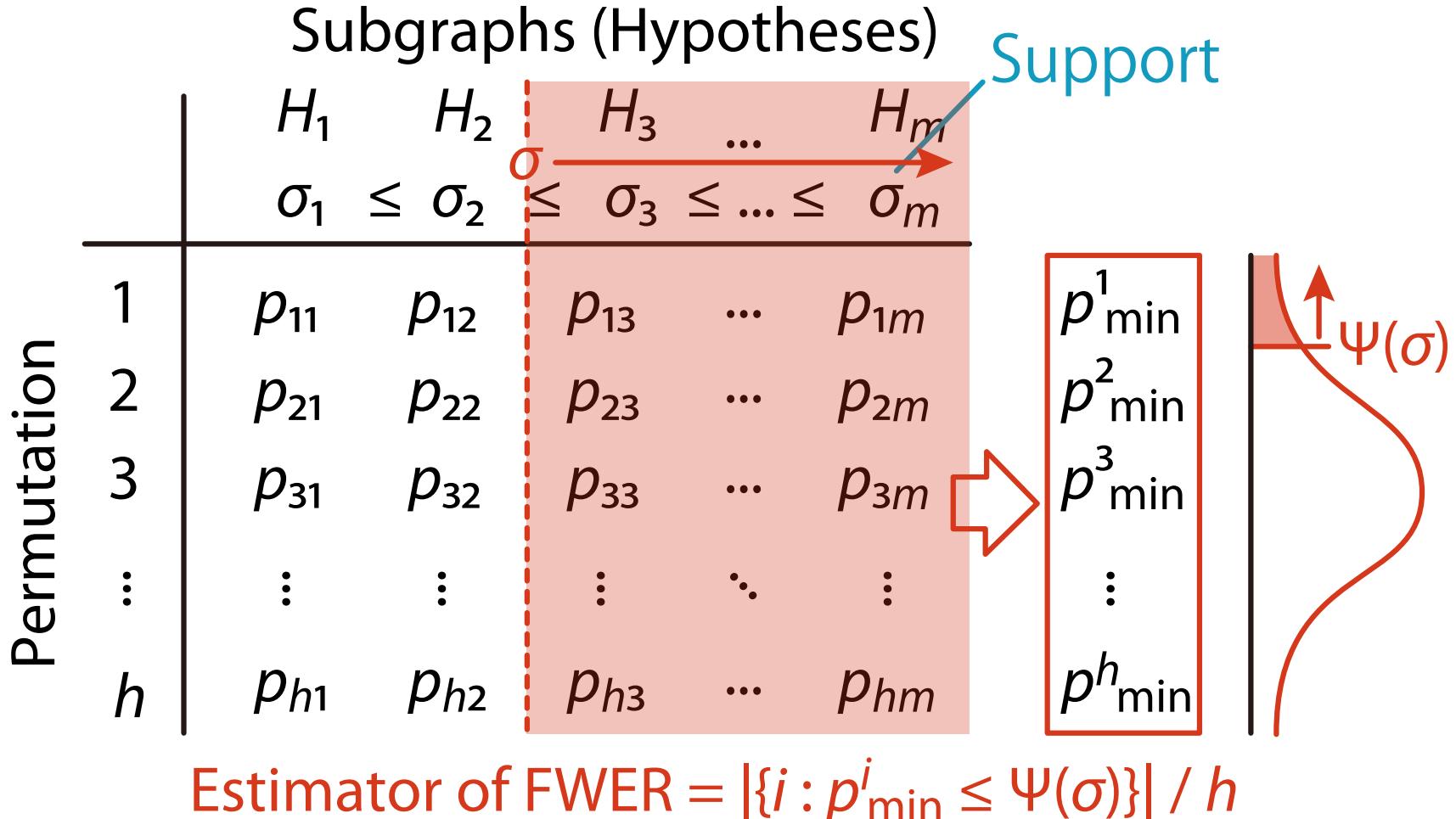
Westfall-Young Permutation



Using Support for Estimating FWER



Estimating FWER

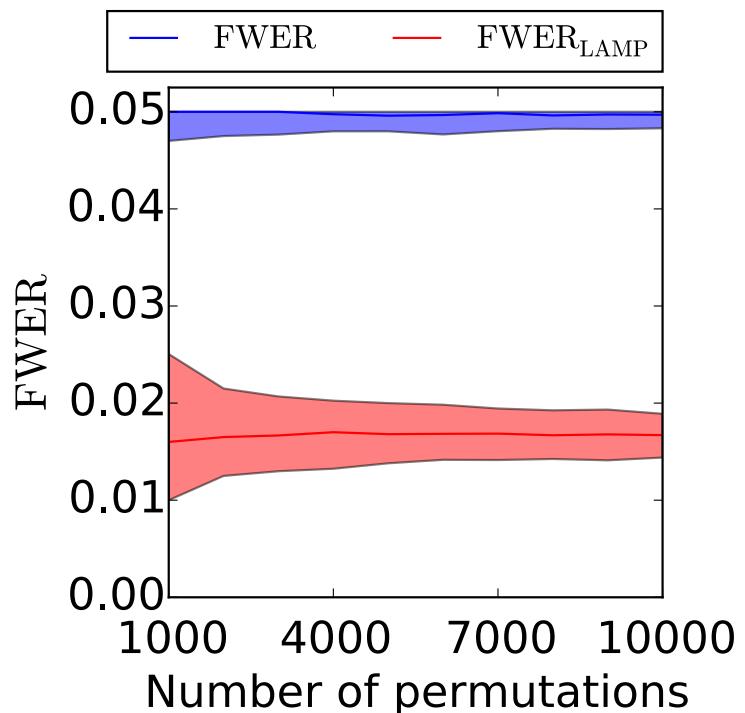


“Westfall-Young light” [Llinares-López et al. KDD’15]

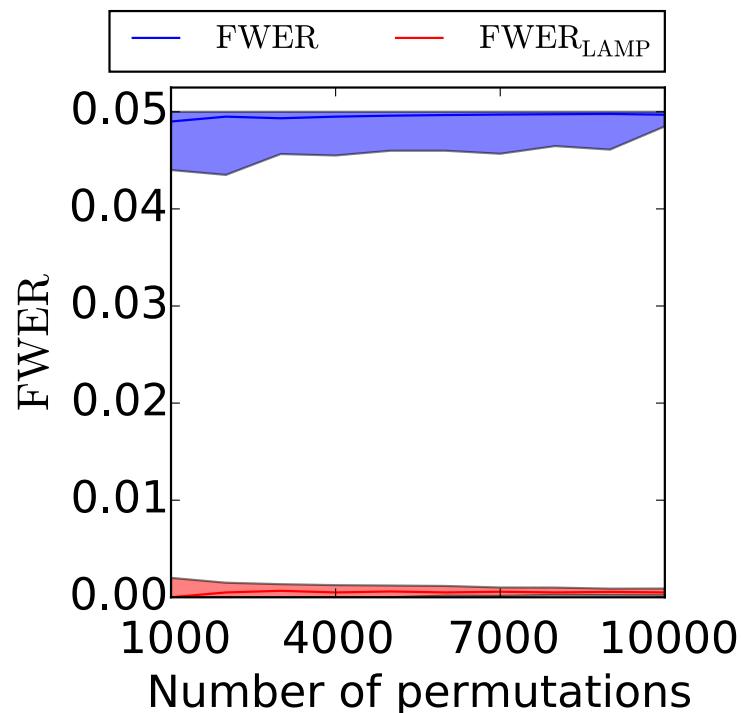
- Precompute h permuted labels; $\sigma \leftarrow 1$; $p_{\min}^i \leftarrow 1$
- Westfall-Young light does the following whenever a miner (like Gaston) finds a new frequent subgraph H :
 - **for** $i \leftarrow 1$ **to** h **do**:
 - $p^i \leftarrow$ the p -value of H for i th permutation
 - $p_{\min}^i \leftarrow \min\{p_{\min}^i, p^i\}$
 - $\text{FWER} \leftarrow |\{i : p_{\min}^i \leq \Psi(\sigma)\}| / h$ // current FWER estimate
 - **while** $\text{FWER} > \alpha$ **do**:
 - $\sigma \leftarrow \sigma + 1$ // σ is the minimum support for mining
 - $\text{FWER} \leftarrow |\{i : p_{\min}^i \leq \Psi(\sigma)\}| / h$
 - Go children of H

FWER in Subgraph Mining

ENZYMES



NCI220



from [Llinares-López et al. KDD2015]

Conclusion

- Significant subgraph mining is introduced
 - Find statistically significant subgraphs while controlling the FWER
 - pattern mining (data mining) + MCP (statistics)
 - Sugiyama, M., Llinares-López, F., Kasenburg, N., Borgwardt, K.: **Significant Subgraph Mining with Multiple Testing Correction**, SIAM SDM 2015
 - Llinares-López, F., Sugiyama, M., Papaxanthos, L., Borgwardt, K.: **Fast and Memory-Efficient Significant Pattern Mining via Permutation Testing**, ACM SIGKDD 2015
- Ongoing projects:
 - Find significant subgraphs on a single massive graph
 - Find significant subtrees on a tree