# **Glossary of Notation**

## **Graph Notation**

G = (V, E)	A graph $G$ , with vertex set $V$ and edge set $E$ .
$N_{v}, N_{e}$	Number of vertices; number of edges.
$\mathscr{G}$	A collection of graphs.
$V^m,V^{(m)}$	Set of ordered/unordered vertex <i>m</i> -tuples.
$V_{\setminus\{i,j\}}^{(m)}$	Set of unordered vertex <i>m</i> -tuples from $V \setminus \{i, j\}$ .
$\mathbf{A} = [A_{ij}]_{i,j=1}^{N_{v}}$	Adjacency matrix.
В	Routing matrix (more generally, incidence matrix).
$\mathbf{L},  ilde{\mathbf{L}}$	Laplacian and normalized Laplacian matrices.
$\{d_{\nu}\}_{\nu\in V}$	Degree sequence.
dist(v, u)	Geodesic distance between <i>v</i> and <i>u</i> .
$\bar{l}(G)$	Average geodesic distance on <i>G</i> .
$c_{Cl}(v), c_B(v), c_{Ei}(v)$	Closeness, betweenness, and eigenvector centrality of $v$ .
den(G)	Density of G.
$ au_3(G)$	Number of connected triples of vertices in <i>G</i> .
$ au_{ riangle}(G)$	Number of triangles in <i>G</i> .
$cl(G), cl_T(G)$	Clustering and clustering-transitivity coefficient of <i>G</i> .
$\mathscr{C} = \{C_1, \ldots, C_K\}$	A partition (e.g., of a set of vertices).
$mod(\mathscr{C})$	Modularity of the partition $\mathscr{C}$ .
$\mathscr{N}(S)$	Neighborhood of vertices in <i>S</i> .
$\mathcal{N}_{v} = \mathcal{N}(\{v\})$	Neighborhood of vertex <i>v</i> .
$\mathscr{N}_{v}^{+} = \mathscr{N}_{v} \cup \{v\}$	Union of vertex <i>v</i> with its neighborhood.
$\eta(G)$	Characteristic of $G$ (e.g., average degree,
	clustering coefficient, etc.).
$G^* = (V^*, E^*)$	Graph obtained by sampling G.
$T = (V_T, E_T)$	A tree.

## **Probability and Statistics Notation**

Probability of the event <i>B</i> .
Random variable; random vector or matrix.
Observation of $X$ or $X$ .
Random vector $\mathbf{X}$ without its $i$ -th entry.
Indicator random variable for the event <i>B</i> .
Cumulative distribution function (CDF), i.e., $F_X(x) = \mathbb{P}(X \le x)$ .
Probability density (or mass) function.
Expected value (mean) of $X$ ; sometimes written $X$ .
Variance of <i>X</i> i.e., $\mathbb{E}((X - X)^2)$
Covariance of <i>X</i> and <i>Y</i> i.e., $\mathbb{E}((X - X)(Y - Y))$
Correlation of <i>X</i> and <i>Y</i> i.e., $Cov(X,Y)/(\mathbb{V}(X)\mathbb{V}(Y))^{1/2}$ .
Sample mean; observed sample mean.
Parameter; estimate of parameter.
Standard error of $\hat{\theta}$ i.e., $se(\hat{\theta}) = \sqrt{\mathbb{V}(\hat{\theta})}$ .
Mean squared error of $\hat{\theta}$ i.e., $MSE(\hat{\theta}) = \mathbb{E}((\hat{\theta} - \theta)^2)$ .
Likelihood function.
Log-likelihood function i.e., $\ell(\theta) = \log \mathcal{L}(\theta)$ .
Residual sum of squares.
False discovery rate.

#### **Other Notation**

n!	$n \times (n-1) \times \cdots \times 2 \times 1$
$\binom{n}{k}$	n!/(k!(n-k)!)
IR	The real numbers.
S	Number of elements in the set <i>S</i> .
$h_n = O(g_n)$	$ h_n/g_n $ is bounded for large n.
NP	A formal characterization in computer science of the difficulty of
	a problem, effectively indicating that it may be solved only
	in a number of computations growing exponentially with
	the size of the problem.
$\mathbf{x}^T$	Transpose of the vector <b>x</b> .

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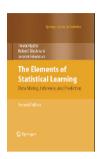
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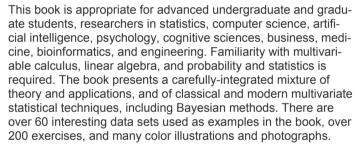
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