
List of Notation

\circ	entry-wise multiplication of vectors
∇	the gradient: $\nabla f(x) = (D_1 f(x), \dots, D_n f(x))$
\ni	$S \ni i$ is equivalent to $i \in S$
\oplus	logical XOR (exclusive-or)
$\ \hat{f}\ _p$	$(\sum_{\gamma \in \mathbb{F}_2^n} \hat{f}(\gamma) ^p)^{1/p}$
Δ	symmetric difference of sets; i.e., $S \Delta T = \{i : i \text{ is in exactly one of } S, T\}$
\vee	logical OR
\wedge	logical AND
$*$	the convolution operator
$[z^k]F(z)$	coefficient on z^k in the power series $F(z)$
1_A	0-1 indicator function for A
1_B	0-1 indicator random variable for event B
2^A	the set of all subsets of A
AND_n	the logical AND function on n bits: False unless all inputs are True
$\chi(b)$	when $b \in \mathbb{F}_2^n$, denotes $(-1)^b \in \mathbb{R}$
$\chi_S(x)$	when $x \in \mathbb{R}^n$, denotes $\prod_{i \in S} x_i$, where $S \subseteq [n]$; when $x \in \mathbb{F}_2^n$, denotes $(-1)^{\sum_{i \in S} x_i}$ if $x \in \mathbb{F}_2^n$
$\text{codim } H$	for subspace $H \leq \mathbb{F}^n$, denotes $n - \dim H$
$\mathbf{Cov}[f, g]$	the covariance of f and g , $\mathbf{Cov}[f] = \mathbf{E}[fg] - \mathbf{E}[f]\mathbf{E}[g]$
$d_{\chi^2}(\varphi, 1)$	chi-squared distance of the distribution with density φ from the uniform distribution

$\deg(f)$	the degree of f ; the least k such that f is a real linear combination of k -juntas
D_i	the i th discrete derivative: $D_i f(x) = \frac{f(x^{(i \rightarrow 1)}) - f(x^{(i \rightarrow -1)})}{2}$
$\text{dist}(g, h)$	the relative Hamming distance; i.e., the fraction of places where g and h disagree
$\text{Dist}(x, y)$	the Hamming distance, $\#\{i : x_i \neq y_i\}$
$\text{DNF}_{\text{size}}(f)$	least possible size of a DNF formula computing f
$\text{DNF}_{\text{width}}(f)$	least possible width of a DNF formula computing f
$\text{DT}(f)$	least possible depth of a decision tree computing f
$\text{DT}_{\text{size}}(f)$	least possible size of a decision tree computing f
$d_{\text{TV}}(\varphi, \psi)$	total variation distance between the distributions with densities φ, ψ
E_i	the i th expectation operator: $E_i f(x) = \mathbf{E}_{x_i}[f(x_1, \dots, x_{i-1}, x_i, x_{i+1}, \dots, x_n)]$
$f \otimes g$	if $f : \{-1, 1\}^m \rightarrow \{-1, 1\}$ and $g : \{-1, 1\}^n \rightarrow \{-1, 1\}$, denotes the function $h : \{-1, 1\}^{mn} \rightarrow \{-1, 1\}$ defined by $h(x^{(1)}, \dots, x^{(m)}) = f(g(x^{(1)}), \dots, g(x^{(m)}))$
$f^{\otimes d}$	if $f : \{-1, 1\}^n \rightarrow \{-1, 1\}$ then $f^{\otimes d} : \{-1, 1\}^{n^d} \rightarrow \{-1, 1\}$ is defined inductively by $f^{\otimes 1} = f$, $f^{\otimes(d+1)} = f \otimes f^{\otimes d}$
f^\dagger	the boolean dual defined by $f^\dagger(x) = -f(-x)$
\mathbb{F}_2	the finite field of size 2
$\widehat{\mathbb{F}}_2^n$	the group (vector space) indexing the Fourier characters of functions $f : \mathbb{F}_2^n \rightarrow \mathbb{R}$
f^{even}	the even part of f , $(f(x) + f(-x))/2$
$\langle f, g \rangle$	$\mathbf{E}_x[f(x)g(x)]$
$\hat{f}(i)$	shorthand for $\hat{f}(\{i\})$ when $i \in \mathbb{N}$
$f_{H z}$	if $f : \mathbb{F}_2^n \rightarrow \mathbb{R}$, $H \leq \mathbb{F}_2^n$, and $z \in H^\perp$, denotes the function $H \rightarrow \mathbb{R}$ defined by $f_{H z}(y) = f(y + z)$
$f _z$	if $f : \{-1, 1\}^n \rightarrow \mathbb{R}$, $J \subseteq [n]$, and $z \in \{-1, 1\}^{\bar{J}}$, denotes the restriction of f given by fixing the coordinates in \bar{J} to z
$f_{J z}$	if $f : \mathbb{F}_2^n \rightarrow \mathbb{R}$, $J \subseteq [n]$, and $z \in \mathbb{F}_2^{\bar{J}}$, denotes the restriction of f given by fixing the coordinates in \bar{J} to z
$f^{=k}$	$\sum_{ S =k} \hat{f}(S) \chi_S$
$f^{\leq k}$	$\sum_{ S \leq k} \hat{f}(S) \chi_S$
f^{odd}	the odd part of f , $(f(x) - f(-x))/2$
$\hat{f}(S)$	the Fourier coefficient of f on character χ_S
$\mathbf{F}_{\beta H^\perp} f(z)$	for $\beta \in H \leq \mathbb{F}_2^n$, denotes $\widehat{f_{H z}}(\beta)$

$F_{S \bar{J}}f(z)$	for $S \subseteq J \subseteq [n]$, denotes $\widehat{f_{J z}}(S)$
$\mathbf{Inf}_i[f]$	the influence of coordinate i on f
$\mathbf{Inf}_i^{(\rho)}[f]$	the noisy influence, $\mathbf{Stab}_\rho[D_i f]$
\bar{J}	if $J \subseteq [n]$, denotes $[n] \setminus J$
L	the Laplacian operator: $Lf = \sum_{i=1}^n L_i f$
$\Lambda_\rho(\mu)$	$\Pr[\mathbf{z}_1 > t, \mathbf{z}_2 > t]$, where $\mathbf{z}_1, \mathbf{z}_2$ are standard Gaussians with correlation $\mathbf{E}[\mathbf{z}_1 \mathbf{z}_2] = \rho$ and $t = \bar{\Phi}^{-1}(\mu)$.
L_i	the i th directional Laplacian operator: $L_i f = f - \mathbf{E}_i f$
$\ln x$	$\log_e x$
$\log x$	$\log_2 x$
Maj_n	the majority function on n bits
$[n]$	$\{1, 2, 3, \dots, n\}$
\mathbb{N}	$\{0, 1, 2, 3, \dots\}$
$N(0, 1)$	the standard Gaussian distribution
\mathbb{N}^+	$\{1, 2, 3, \dots\}$
$N_\rho(x)$	the probability distribution generating strings ρ -correlated to x
$\mathbf{NS}_\delta[f]$	the noise sensitivity of f at δ ; i.e., $\frac{1}{2} - \frac{1}{2} \mathbf{Stab}_{1-2\delta}[f]$
OR_n	the logical OR function on n bits: True unless all inputs are False
Φ	the standard Gaussian cdf, $\Phi(t) = \int_{-\infty}^t \phi(z) dz$
ϕ	the standard Gaussian pdf, $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-z^2/2}$
φ_A	the density function for the uniform probability distribution on A ; i.e., $1_A/\mathbf{E}[1_A]$
$\bar{\Phi}$	the standard Gaussian complementary cdf, $\bar{\Phi}(t) = \int_t^\infty \phi(z) dz$
\mathbb{R}	the real numbers
$\mathbb{R}^{\geq 0}$	the nonnegative real numbers
$\text{sens}_f(x)$	the number of pivotal coordinates for f at x
$\text{sgn}(t)$	+1 if $t \geq 0$, -1 if $t < 0$
S_n	the symmetric group on $[n]$
$\text{sparsity}(f)$	$\Pr_{\mathbf{x}}[f(\mathbf{x}) \neq 0]$
$\text{sparsity}(\hat{f})$	$ \text{supp}(\hat{f}) $
$\mathbf{Stab}_\rho[f]$	the noise stability of f at ρ : $\mathbf{E}[f(\mathbf{x})f(\mathbf{y})]$ where \mathbf{x}, \mathbf{y} are a ρ -correlated pair
$\text{supp}(f)$	the set of inputs where f is nonzero

T_ρ	the noise operator: $T_\rho f(x) = \mathbf{E}_{\mathbf{y} \sim N_\rho(x)}[f(\mathbf{y})]$
\mathcal{U}	the Gaussian isoperimetric function, $\mathcal{U} = \phi \circ \Phi^{-1}$
$\mathbf{Var}[f]$	the variance of f , $\mathbf{Var}[f] = \mathbf{E}[f^2] - \mathbf{E}[f]^2$
Var_i	defined by $\text{Var}_i f(x) = \mathbf{Var}_{\mathbf{x}_i}[f(x_1, \dots, x_{i-1}, \mathbf{x}_i, x_{i+1}, \dots, x_n)]$
$\mathbf{W}^k[f]$	the Fourier weight of f at degree k
$\mathbf{W}^{>k}[f]$	the Fourier weight of f at degrees above k
$\mathbf{x} \sim \{-1, 1\}^n$	the random variable \mathbf{x} is chosen uniformly from $\{-1, 1\}^n$
$\mathbf{x} \sim A$	the random variable \mathbf{x} is chosen uniformly from the set A
$x^{(i \mapsto b)}$	the string $(x_1, \dots, x_{i-1}, b, x_{i+1}, \dots, x_n)$
$x^{\oplus i}$	$(x_1, \dots, x_{i-1}, -x_i, x_{i+1}, \dots, x_n)$
$\mathbf{x} \sim \varphi$	the random variable \mathbf{x} is chosen from the probability distribution with density φ
x^S	$\prod_{i \in S} x_i$, with the convention $x^\emptyset = 1$
(y, z)	if $J \subseteq [n]$, $y \in \{-1, 1\}^J$, $z \in \{-1, 1\}^{\bar{J}}$, denotes the natural composite string in $\{-1, 1\}^n$