

Notation

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1 Font notation

| | |
|---|----------|
| $a, b, c, \dots, A, B, C, \dots$ | Scalars |
| $\mathbf{a}, \mathbf{b}, \mathbf{c}, \dots$ | Vectors |
| $\mathbf{A}, \mathbf{B}, \mathbf{C}, \dots$ | Matrices |
| $\mathcal{A}, \mathcal{B}, \mathcal{C}, \dots$ | Tensors |
| $A, B, C, \dots, \mathcal{A}, \mathcal{B}, \mathcal{C}, \dots, \mathbb{A}, \mathbb{B}, \mathbb{C}, \dots$ | Sets |

2 Signals and functions

2.1 Time indexing

| | |
|---------------------------------|--|
| $x(t)$ | Continuous-time t |
| $x[n], x[k], x[m], x[i], \dots$ | Discrete-time n, k, m, i, \dots (parenthesis should be adopted only if there are no continuous-time signals in the context to avoid ambiguity) |
| $x_n, x_k, x_m, x_i, \dots$ | |
| $x(n), x(k), x(m), x(i), \dots$ | |
| $x[((n-m))_N], x((n-m))_N$ | Circular shift in m samples within a N -samples window [10, 14] |

2.2 Common functions

| | |
|---------------------------|---|
| $\delta(t)$ | Delta function |
| $\delta[n], \delta_{i,j}$ | Kronecker function ($n = i - j$) |
| $h(t), h[n]$ | Impulse response (continuous and discrete time) |

| | |
|------------------------------|--|
| $\tilde{x}[n], \tilde{x}(t)$ | Periodic discrete- or continuous-time signal |
| $\hat{x}[n], \hat{x}(t)$ | Estimate of $x[n]$ or $x(t)$ |
| $\dot{x}[m]$ | Interpolation of $x[n]$ |

2.3 Operations and symbols

| | |
|---|---|
| $f : A \rightarrow B$ | A function f whose domain is A and codomain is B |
| $\mathbf{f} : A \rightarrow \mathbb{R}^n$ | A vector-valued function \mathbf{f} , i.e., $n \geq 2$ |
| $f^n, x^n(t), x^n[k]$ | n th power of the function f , $x[n]$ or $x(t)$ |
| $f^{(n)}, x^{(n)}(t)$ | n th derivative of the function f or $x(t)$ |
| $f', f^{(1)}, x'(t)$ | 1th derivative of the function f or $x(t)$ |
| $f'', f^{(2)}, x''(t)$ | 2th derivative of the function f or $x(t)$ |
| $\arg \max_{x \in \mathcal{A}} f(x)$ | Value of x that minimizes x |
| $\arg \min_{x \in \mathcal{A}} f(x)$ | Value of x that minimizes x |
| $f(\mathbf{x}) = \inf_{\mathbf{y} \in \mathcal{A}} g(\mathbf{x}, \mathbf{y})$ | Infimum, i.e., $f(\mathbf{x}) = \min \{g(\mathbf{x}, \mathbf{y}) \mid \mathbf{y} \in \mathcal{A} \wedge (\mathbf{x}, \mathbf{y}) \in \text{dom}(g)\}$, which is the greatest lower bound of this set [2] |
| $f(\mathbf{x}) = \sup_{\mathbf{y} \in \mathcal{A}} g(\mathbf{x}, \mathbf{y})$ | Supremum, i.e., $f(\mathbf{x}) = \max \{g(\mathbf{x}, \mathbf{y}) \mid \mathbf{y} \in \mathcal{A} \wedge (\mathbf{x}, \mathbf{y}) \in \text{dom}(g)\}$, which is the least upper bound of this set [2] |
| $f \circ g$ | Composition of the functions f and g |
| $*$ | Convolution (discrete or continuous) |
| $\otimes, \textcircled{\mathbb{N}}$ | Circular convolution [6, 14] |

2.4 Transformations

| | |
|--------------------------|---|
| W_N | Twiddle factor, $e^{-j\frac{2\pi}{N}}$ [10] |
| $\mathcal{F}\{\cdot\}$ | Fourier transform |
| $\mathcal{L}\{\cdot\}$ | Laplace transform |
| $\mathcal{Z}\{\cdot\}$ | z-transform |
| $\hat{x}(t), \hat{x}[n]$ | Hilbert transform of $x(t)$ or $x[n]$ |
| $X(s)$ | Laplace transform of $x(t)$ |

| | |
|---|--|
| $X(f)$ | Fourier transform (FT) (in linear frequency, Hz) of $x(t)$ |
| $X(j\omega)$ | Fourier transform (FT) (in angular frequency, rad/sec) of $x(t)$ |
| $X(e^{j\omega})$ | Discrete-time Fourier transform (DTFT) of $x[n]$ |
| $X[k], X(k), X_k$ | Discrete Fourier transform (DFT) or fast Fourier transform (FFT) of $x[n]$, or even the Fourier series (FS) of the periodic signal $x(t)$ |
| $\tilde{X}[k], \tilde{X}(k), \tilde{X}_k$ | Discrete Fourier series (DFS) of $\tilde{x}[n]$ |
| $X(z)$ | z -transform of $x[n]$ |

3 Probability, statistics, and stochastic processes

3.1 Operators and symbols

| | |
|--|--|
| $\mathbb{E}[\cdot], E[\cdot], \mathbb{E}[\cdot]$ | Statistical expectation operator [5, 13] |
| $\mathbb{E}_u[\cdot], E_u[\cdot], \mathbb{E}_u[\cdot]$ | Statistical expectation operator with respect to u |
| $\langle \cdot \rangle$ | Ensamble average |
| $\text{var}[\cdot], \text{VAR}[\cdot]$ | Variance operator [1, 9, 12, 16] |
| $\text{var}_u[\cdot][\cdot], \text{VAR}_u[\cdot]$ | Variance operator with respect to u |
| $\text{cov}[\cdot], \text{COV}[\cdot]$ | Covariance operator [1] |
| $\text{cov}_u[\cdot], \text{COV}_u[\cdot]$ | Covariance operator with respect to u |
| μ_x | Mean of the random variable x |
| $\boldsymbol{\mu}_x, \mathbf{m}_x$ | Mean vector of the random variable \mathbf{x} [3] |
| μ_n | n th-order moment of a random variable |
| σ_x^2, κ_2 | Variance of the random variable x |
| \mathcal{K}_x, μ_4 | Kurtosis (4th-order moment) of the random variable x |
| κ_n | n th-order cumulant of a random variable |
| $\rho_{x,y}$ | Pearson correlation coefficient between x and y |
| $a \sim P$ | Random variable a with distribution P |
| \mathcal{R} | Rayleigh's quotient |

3.2 Stochastic processes

| | |
|---|--|
| $r_x(\tau), R_x(\tau)$ | Autocorrelation function of the signal $x(t)$ or $x[n]$ [13] |
| $S_x(f), S_x(j\omega)$ | Power spectral density (PSD) of $x(t)$ in linear (f) or angular (ω) frequency |
| $S_{x,y}(f), S_{x,y}(j\omega)$ | Cross PSD of $x(t)$ and $y(t)$ in linear or angular (ω) frequency |
| \mathbf{R}_x | (Auto)correlation matrix of $\mathbf{x}(n)$ |
| $r_{x,d}(\tau), R_{x,d}(\tau)$ | Cross-correlation between $x[n]$ and $d[n]$ or $x(t)$ and $d(t)$ [13] |
| \mathbf{R}_{xy} | Cross-correlation matrix of $\mathbf{x}(n)$ and $\mathbf{y}(n)$ |
| \mathbf{p}_{xd} | Cross-correlation vector between $\mathbf{x}(n)$ and $d(n)$ [dinizAdaptiveFiltering1997] |
| $c_x(\tau), C_x(\tau)$ | Autocovariance function of the signal $x(t)$ or $x[n]$ [13] |
| $\mathbf{C}_x, \mathbf{K}_x, \mathbf{\Sigma}_x, \text{cov}[\mathbf{x}]$ | (Auto)covariance matrix of \mathbf{x} [9, 12, 16, 19] |
| $c_{xy}(\tau), C_{xy}(\tau)$ | Cross-covariance function of the signal $x(t)$ or $x[n]$ [13] |
| $\mathbf{C}_{xy}, \mathbf{K}_{xy}, \mathbf{\Sigma}_{xy}$ | Cross-covariance matrix of \mathbf{x} and \mathbf{y} |

3.3 Functions

| | |
|--|--|
| $Q(\cdot)$ | Q -function, i.e., $P[\mathcal{N}(0, 1) > x]$ [16] |
| $\text{erf}(\cdot)$ | Error function [16] |
| $\text{erfc}(\cdot)$ | Complementary error function i.e., $\text{erfc}(x) = 2Q(\sqrt{2}x) - \text{erf}(x)$ [16] |
| $P[A]$ | Probability of the event or set A [12] |
| $p(\cdot), f(\cdot)$ | Probability density function (PDF) or probability mass function (PMF) [12] |
| $p(x A)$ | Conditional PDF or PMF [12] |
| $F(\cdot)$ | Cumulative distribution function (CDF) |
| $\Phi_x(\omega), M_x(j\omega), E[e^{j\omega x}]$ | First characteristic function (CF) of x [theodoridisMachineLearningBayesian2020a, 16] |

| | |
|--|---|
| $M_x(t), \Phi_x(-jt), E[e^{tx}]$ | Moment-generating function (MGF) of x [theodoridisMachineLearningBayesian2020a, 16] |
| $\Psi_x(\omega), \ln \Phi_x(\omega), \ln E[e^{j\omega x}]$ | Second characteristic function |
| $K_x(t), \ln E[e^{tx}], \ln M_x(t)$ | Cumulant-generating function (CGF) of x [9] |

3.4 Distributions

| | |
|---|--|
| $\mathcal{N}(\mu, \sigma^2)$ | Gaussian distribution of a random variable with mean μ and variance σ^2 . The same notation can be used to denote a real-valued white Gaussian process with mean equal to μ and power spectral density equal to $N_0/2$, e.g., $s(t) \sim \mathcal{N}(\mu, N_0/2)$ |
| $\mathcal{CN}(\mu, \sigma^2)$ | Complex Gaussian distribution of a random variable with mean μ and variance σ^2 . The same notation can be used to denote a complex-valued white Gaussian process with mean equal to μ and power spectral density equal to N_0 , e.g., $s(t) \sim \mathcal{CN}(\mu, N_0)$ |
| $\mathcal{N}(\mathbf{\mu}, \mathbf{\Sigma})$ | Gaussian distribution of a vector random variable with mean $\mathbf{\mu}$ and covariance matrix $\mathbf{\Sigma}$ |
| $\mathcal{CN}(\mathbf{\mu}, \mathbf{\Sigma})$ | Complex Gaussian distribution of a vector random variable with mean $\mathbf{\mu}$ and covariance matrix $\mathbf{\Sigma}$ |
| $\mathcal{U}(a, b)$ | Uniform distribution from a to b |
| $\chi^2(n), \chi_n^2$ | Chi-square distribution with n degree of freedom (assuming that the Gaussians are $\mathcal{N}(0, 1)$) |
| $\text{Exp}(\lambda)$ | Exponential distribution with rate parameter λ |
| $\Gamma(\alpha, \beta)$ | Gamma distribution with shape parameter α and rate parameter β |
| $\Gamma(\alpha, \theta)$ | Gamma distribution with shape parameter α and scale parameter $\theta = 1/\beta$ |
| Nakagami(m, Ω) | Nakagami-m distribution with shape parameter m and spread parameter Ω |
| Rayleigh(σ) | Rayleigh distribution with scale parameter σ |

| | |
|----------------------|--|
| Rayleigh(Ω) | Rayleigh distribution with the second moment $\Omega = E[x^2] = 2\sigma^2$ |
| Rice(s, σ) | Rice distribution with noncentrality parameter (specular component) s and σ |
| Rice(A, K) | Rice distribution with Rice factor $K = s^2/2\sigma^2$ and scale parameter $A = s^2 + 2\sigma^2$ |

4 Statistical signal processing

| | |
|--|--|
| $\nabla f, \mathbf{g}$ | Gradient descent vector |
| $\nabla_x f, \mathbf{g}_x$ | Gradient descent vector with respect x |
| \mathbf{g} (or $\hat{\mathbf{g}}$ if the gradient vector is \mathbf{g}) | Stochastic gradient descent (SGD) |
| $J(\cdot), \mathcal{E}(\cdot)$ | Cost-function or objective function |
| $\Lambda(\cdot)$ | Likelihood function |
| $\Lambda_l(\cdot)$ | Log-likelihood function |
| $\hat{x}(t)$ or $\hat{x}[n]$ | Estimate of $x(t)$ or $x[n]$ |
| $\hat{\mu}_x, \hat{\mathbf{m}}_x$ | Sample mean of $x[n]$ or $x(t)$ |
| $\hat{\mu}_{\mathbf{x}}, \hat{\mathbf{m}}_{\mathbf{x}}$ | Sample mean vector of $\mathbf{x}[n]$ or $\mathbf{x}(t)$ |
| $\hat{r}_x(\tau), \hat{R}_x(\tau)$ | Estimated autocorrelation function of the signal $x(t)$ or $x[n]$ |
| $\hat{S}_x(f), \hat{S}_x(j\omega)$ | Estimated power spectral density (PSD) of $x(t)$ in linear (f) or angular (ω) frequency |
| $\hat{\mathbf{R}}_x$ | Sample (auto)correlation matrix |
| $\hat{r}_{x,d}(\tau), \hat{R}_{x,d}(\tau)$ | Estimated cross-correlation between $x[n]$ and $d[n]$ or $x(t)$ and $d(t)$ |
| $\hat{S}_{x,y}(f), \hat{S}_{x,y}(j\omega)$ | Estimated cross PSD of $x(t)$ and $y(t)$ in linear or angular (ω) frequency |
| $\hat{\mathbf{R}}_{xy}$ | Sample cross-correlation matrix of \mathbf{R}_{xy} |
| $\hat{\rho}_{x,y}$ | Estimated Pearson correlation coefficient between x and y |
| $\hat{c}_x(\tau), \hat{C}_x(\tau)$ | Estimated autocovariance function of the signal $x(t)$ or $x[n]$ |
| $\hat{\mathbf{C}}_x, \hat{\mathbf{K}}_x, \hat{\Sigma}_x$ | Sample (auto)covariance matrix |
| $\hat{c}_{xy}(\tau), \hat{C}_{xy}(\tau)$ | Estimated cross-covariance function of the signal $x(t)$ or $x[n]$ |
| $\hat{\mathbf{C}}_{xy}, \hat{\mathbf{K}}_{xy}, \hat{\Sigma}_{xy}$ | Sample cross-covariance matrix |
| $\mathbf{w}, \boldsymbol{\theta}$ | Parameters, coefficients, or weights vector |

| | |
|--|--|
| $\mathbf{w}_o, \mathbf{w}^*, \boldsymbol{\theta}_o, \boldsymbol{\theta}^*$ | Optimum value of the parameters, coefficients, or weights vector |
| \mathbf{W} | Matrix of the weights |
| \mathbf{J} | Jacobian matrix |
| \mathbf{H} | Hessian matrix |
| $\hat{\mathbf{H}}$ | Estimate of the Hessian matrix |

5 Linear Algebra

5.1 Common matrices and vectors

| | |
|--|---|
| \mathbf{W}, \mathbf{D} | Diagonal matrix |
| \mathbf{P} | Projection matrix; Permutation matrix |
| \mathbf{J} | Jordan matrix |
| \mathbf{L} | Lower matrix |
| \mathbf{U} | Upper matrix |
| \mathbf{C} | Cofactor matrix |
| $\mathbf{C}_A, \text{cof}(\mathbf{A})$ | Cofactor matrix of \mathbf{A} |
| \mathbf{S} | Symmetric matrix |
| \mathbf{Q} | Orthogonal matrix |
| \mathbf{I}_N | $N \times N$ -dimensional identity matrix |
| $\mathbf{0}_{M \times N}$ | $M \times N$ -dimensional null matrix |
| $\mathbf{0}_N$ | N -dimensional null vector |
| $\mathbf{1}_{M \times N}$ | $M \times N$ -dimensional ones matrix |
| $\mathbf{1}_N$ | N -dimensional ones vector |
| $\mathbf{0}$ | Null matrix, vector, or tensor (dimensionality understood by context) |
| $\mathbf{1}$ | Ones matrix, vector, or tensor (dimensionality understood by context) |

5.2 Indexing

| | |
|--|--|
| $x_{i_1, i_2, \dots, i_N}, [\mathcal{X}]_{i_1, i_2, \dots, i_N}$ | Element in the position (i_1, i_2, \dots, i_N) of the tensor \mathcal{X} |
| $\mathcal{X}^{(n)}$ | n th tensor of a nontemporal sequence |
| $\mathbf{x}_n, \mathbf{x}_{:n}$ | n th column of the matrix \mathbf{X} |
| $\mathbf{x}_{n:}$ | n th row of the matrix \mathbf{X} |
| $\mathbf{x}_{i_1, \dots, i_{n-1}, :, i_{n+1}, \dots, i_N}$ | Mode- n fiber of the tensor \mathcal{X} |
| $\mathbf{x}_{:, i_2, i_3}$ | Column fiber (mode-1 fiber) of the thrid-order tensor \mathcal{X} |
| $\mathbf{x}_{i_1, :, i_3}$ | Row fiber (mode-2 fiber) of the thrid-order tensor \mathcal{X} |

| | |
|--|---|
| $\mathbf{x}_{i_1, i_2, :}$ | Tube fiber (mode-3 fiber) of the thrid-order tensor \mathcal{X} |
| $\mathbf{X}_{i_1, :, :}$ | Horizontal slice of the thrid-order tensor \mathcal{X} |
| $\mathbf{X}_{:, i_2, :}$ | Lateral slices slice of the thrid-order tensor \mathcal{X} |
| $\mathbf{X}_{i_3}, \mathbf{X}_{:, :, i_3}$ | Frontal slices slice of the thrid-order tensor \mathcal{X} |

5.3 General operations

| | |
|--------------------------------|--|
| $\langle \cdot, \cdot \rangle$ | Inner product, e.g., $\langle \mathbf{a}, \mathbf{b} \rangle = \mathbf{a}^\top \mathbf{b}$ |
| \circ | Outer product, e.g., $\mathbf{a} \circ \mathbf{b} = \mathbf{a} \mathbf{b}^\top$ |
| \otimes | Kronecker product |
| \odot | Hadamard (or Schur) (elementwise) product |
| $\cdot^{\odot n}$ | n th-order Hadamard power |
| $\cdot^{\odot \frac{1}{n}}$ | n th-order Hadamard root |
| \oslash | Hadamard (or Schur) (elementwise) division |
| \diamond | Khatri-Rao product |
| \otimes | Kronecker Product |
| \times_n | n -mode product |

5.4 Operations with matrices and tensors

| | |
|------------------------------------|---|
| \mathbf{A}^{-1} | Inverse matrix |
| $\mathbf{A}^+, \mathbf{A}^\dagger$ | Moore-Penrose left pseudoinverse |
| \mathbf{A}^\top | Transpose |
| $\mathbf{A}^{-\top}$ | Transpose of the inverse, i.e., $(\mathbf{A}^{-1})^\top = (\mathbf{A}^\top)^{-1}$ [7, 15] |
| \mathbf{A}^* | Complex conjugate |
| \mathbf{A}^H | Hermitian |
| $\ \mathbf{A}\ _F$ | Frobenius norm |
| $\ \mathbf{A}\ $ | Matrix norm |
| $ \mathbf{A} , \det(\mathbf{A})$ | Determinant |
| $\text{diag}(\mathbf{A})$ | The elements in the diagonal of \mathbf{A} |
| $\mathbf{E}[\mathbf{A}]$ | Vectorization: stacks the columns of the matrix \mathbf{A} into a long column vector |
| $\mathbf{E}_d[\mathbf{A}]$ | Extracts the diagonal elements of a square matrix and returns them in a column vector |

| | |
|-----------------------------|---|
| $\mathbf{E}_l [\mathbf{A}]$ | Extracts the elements strictly below the main diagonal of a square matrix in a column-wise manner and returns them into a column vector |
| $\mathbf{E}_u [\mathbf{A}]$ | Extracts the elements strictly above the main diagonal of a square matrix in a column-wise manner and returns them into a column vector |
| $\mathbf{E}_b [\mathbf{A}]$ | Block vectorization operator: stacks square block matrices of the input into a long block column matrix |
| $\text{unvec}(\mathbf{A})$ | Reshapes a column vector into a matrix |
| $\text{tr}\{\mathbf{A}\}$ | trace |
| $\mathbf{X}_{(n)}$ | n -mode matricization of the tensor \mathcal{X} |

5.5 Operations with vectors

| | |
|------------------------------------|--|
| $\ \mathbf{a}\ $ | l_1 norm, 1-norm, or Manhattan norm |
| $\ \mathbf{a}\ , \ \mathbf{a}\ _2$ | l_2 norm, 2-norm, or Euclidean norm |
| $\ \mathbf{a}\ _p$ | l_p norm, p -norm, or Minkowski norm |
| $\ \mathbf{a}\ _\infty$ | l_∞ norm, ∞ -norm, or Chebyshev norm |
| $\text{diag}(\mathbf{a})$ | Diagonalization: a square, diagonal matrix with entries given by the vector \mathbf{a} |

5.6 Decompositions

| | |
|-----------------------|--|
| $\mathbf{\Lambda}$ | Eigenvalue matrix [18] |
| \mathbf{Q} | Eigenvectors matrix; Orthogonal matrix of the QR decomposition[18] |
| \mathbf{R} | Upper triangular matrix of the QR decomposition[18] |
| \mathbf{U} | Left singular vectors[18] |
| \mathbf{U}_r | Left singular nondegenerated vectors |
| $\mathbf{\Sigma}$ | Singular value matrix |
| $\mathbf{\Sigma}_r$ | Singular value matrix with nonzero singular values in the main diagonal |
| $\mathbf{\Sigma}^+$ | Singular value matrix of the pseudoinverse [18] |
| $\mathbf{\Sigma}_r^+$ | Singular value matrix of the pseudoinverse with nonzero singular values in the main diagonal |

| | |
|---|---|
| \mathbf{V} | Right singular vectors [18] |
| \mathbf{V}_r | Right singular nondegenerated vectors |
| $\text{eig}(\mathbf{A})$ | Set of the eigenvalues of \mathbf{A} [4, 12, 15] |
| $\llbracket \mathbf{A}, \mathbf{B}, \mathbf{C}, \dots \rrbracket$ | CANDECOMP/PARAFAC (CP) decomposition of the tensor \mathcal{X} from the outer product of column vectors of $\mathbf{A}, \mathbf{B}, \mathbf{C}, \dots$ |
| $\llbracket \boldsymbol{\lambda}; \mathbf{A}, \mathbf{B}, \mathbf{C}, \dots \rrbracket$ | Normalized CANDECOMP/PARAFAC (CP) decomposition of the tensor \mathcal{X} from the outer product of column vectors of $\mathbf{A}, \mathbf{B}, \mathbf{C}, \dots$ |

5.7 Spaces

| | |
|---|--|
| $\text{span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\}$ | Vector space spanned by the argument vectors [7] |
| $\mathbf{C}(\mathbf{A}), \text{columnspace}(\mathbf{A}), \text{range}(\mathbf{A}), \text{span}\{\mathbf{A}\}, \text{image}(\mathbf{A})$ | Columnspace, range or image, i.e., the space $\text{span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\}$, where \mathbf{a}_i is the i th column vector of the matrix \mathbf{A} [13, 18] |
| $\mathbf{C}(\mathbf{A}^H)$ | Row space (also called left column space) [13, 18] |
| $\mathbf{N}(\mathbf{A}), \text{nullspace}(\mathbf{A}), \text{kernel}(\mathbf{A})$ | Nullspace (or kernel space) [13, 18] |
| $\mathbf{N}(\mathbf{A}^H)$ | Left nullspace |
| $\text{rank } \mathbf{A}$ | Rank, that is, $\dim(\text{span}\{\mathbf{A}\}) = \dim(\mathbf{C}(\mathbf{A}))$ [13] |
| $\text{nullity}(\mathbf{A})$ | Nullity of \mathbf{A} , i.e., $\dim(\mathbf{N}(\mathbf{A}))$ |
| $\mathbf{a} \perp \mathbf{b}$ | \mathbf{a} is orthogonal to \mathbf{b} |
| $\mathbf{a} \not\perp \mathbf{b}$ | \mathbf{a} is not orthogonal to \mathbf{b} |

5.8 Inequalities

| | |
|-----------------------------------|---|
| $\mathcal{X} \leq 0$ | Nonnegative tensor |
| $\mathbf{a} \preceq_K \mathbf{b}$ | Generalized inequality meaning that $\mathbf{b} - \mathbf{a}$ belongs to the conic subset K in the space $\mathbb{R}^n[2]$ |
| $\mathbf{a} \prec_K \mathbf{b}$ | Strict generalized inequality meaning that $\mathbf{b} - \mathbf{a}$ belongs to the interior of the conic subset K in the space $\mathbb{R}^n[2]$ |

| | |
|--------------------------------|--|
| $\mathbf{a} \leq \mathbf{b}$ | Generalized inequality meaning that $\mathbf{b} - \mathbf{a}$ belongs to the nonnegative orthant conic subset, \mathbb{R}_+^n , in the space \mathbb{R}^n . [2] |
| $\mathbf{a} < \mathbf{b}$ | Strict generalized inequality meaning that $\mathbf{b} - \mathbf{a}$ belongs to the positive orthant conic subset, \mathbb{R}_{++}^n , in the space \mathbb{R}^n [2] |
| $\mathbf{A} \leq_K \mathbf{B}$ | Generalized inequality meaning that $\mathbf{B} - \mathbf{A}$ belongs to the conic subset K in the space \mathbb{S}^n [2] |
| $\mathbf{A} <_K \mathbf{B}$ | Strict generalized inequality meaning that $\mathbf{B} - \mathbf{A}$ belongs to the interior of the conic subset K in the space \mathbb{S}^n [2] |
| $\mathbf{A} \leq \mathbf{B}$ | Generalized inequality meaning that $\mathbf{B} - \mathbf{A}$ belongs to the positive semidefinite conic subset, \mathbb{S}_+^n , in the space \mathbb{S}^n [2] |
| $\mathbf{A} < \mathbf{B}$ | Strict generalized inequality meaning that $\mathbf{B} - \mathbf{A}$ belongs to the positive orthant conic subset, \mathbb{S}_{++}^n , in the space \mathbb{S}^n [2] |

6 Communication systems

| | |
|-------------------|---|
| B | One-sided bandwidth of the transmitted signal, in Hz |
| W | One-sided bandwidth of the transmitted signal, in rad/s |
| x_i | Real or in-phase part of x |
| x_q | Imaginary or quadrature part of x |
| f_c, f_{RF} | Carrier frequency (in Hertz) |
| f_L | Carrier frequency in L-band (in Hertz) |
| f_{IF} | Intermediate frequency (in Hertz) |
| f_s | Sampling frequency or sampling rate (in Hertz) |
| T_s | Sampling time interval/duration/period |
| R | Bit rate |
| T | Bit interval/duration/period |
| T_c | Chip interval/duration/period |
| T_{sy}, T_{sym} | Symbol/signaling[16] interval/duration/period |

| | |
|---------------------|---|
| s_{RF} | Transmitted signal in RF |
| s_{FI} | Transmitted signal in FI |
| s, s_l | Lowpass (or baseband) equivalent signal or envelope complex of transmitted signal |
| r_{RF} | Received signal in RF |
| r_{FI} | Received signal in FI |
| r, r_l | Lowpass (or baseband) equivalent signal or envelope complex of received signal |
| ϕ | Signal phase |
| ϕ_0 | Initial phase |
| η_{RF}, w_{RF} | Noise in RF |
| η_{FI}, w_{FI} | Noise in FI |
| η, w | Noise in baseband |
| τ | Timing delay |
| $\Delta\tau$ | Timing error (delay - estimated) |
| φ | Phase offset |
| $\Delta\varphi$ | Phase error (offset - estimated) |
| f_d | Linear Doppler frequency |
| Δf_d | Frequency error (Doppler frequency - estimated) |
| ν | Angular Doppler frequency |
| $\Delta\nu$ | Frequency error (Doppler frequency - estimated) |
| γ, A | Transmitted signal amplitude |
| γ_0, A_0 | Combined effect of the path loss and antenna gain |

7 Discrete mathematics

7.1 Set theory

| | |
|------------------------|--|
| $A + B$ | Set addition (Minkowski sum), i.e., $\{\mathbf{v} \in \mathbb{R}^n \mid \mathbf{v} = \mathbf{x} + \mathbf{y}, \forall \mathbf{x} \in X \wedge \mathbf{y} \in Y\}$ [11] |
| $A - B$ | Minkowski difference, i.e., $\{\mathbf{v} \in \mathbb{R}^n \mid \mathbf{v} = \mathbf{x} - \mathbf{y}, \forall \mathbf{x} \in X \wedge \mathbf{y} \in Y\}$ |
| $A \ominus B$ | Pontryagin difference, i.e., $\{\mathbf{v} \in \mathbb{R}^n \mid \mathbf{v} + \mathbf{y} \in X, \forall \mathbf{y} \in Y\}$ [11] |
| $A \setminus B, A - B$ | Set difference or set subtraction, i.e., $A \setminus B = \{x \mid x \in A \wedge x \notin B\}$ the set containing the elements of A that are not in B [17] |

| | |
|--|---|
| $A \cup B$ | Set of union |
| $A \cap B$ | Set of intersection |
| $A \times B$ | Cartesian product |
| A^n | $\underbrace{A \times A \times \cdots \times A}_{n \text{ times}}$ |
| A^\perp | Orthogonal complement of A , e.g., $N(\mathbf{A}) = C(\mathbf{A}^\top)^\perp$ [2] |
| $A \oplus B$ | Direct sum, i.e., each $\mathbf{v} \in \{\sum \mathbf{a}_i \mid \mathbf{a}_i \in S_i, i = 1, \dots, k\}$ has a unique representation of $\sum \mathbf{a}_i$ with $\mathbf{a}_i \in S_i$. That is, they expand to a space. Note that $\{S_i\}$ might not be orthogonal each other [7] |
| $A \overset{\perp}{\oplus} B$ | Direct sum of two space that are orthogonal and span a n -dimensional space, e.g., $C(\mathbf{A}^\top) \overset{\perp}{\oplus} C(\mathbf{A}^\top)^\perp = \mathbb{R}^n$ (this decomposition of \mathbb{R}^n is called the orthogonal decomposition induced by \mathbf{A}) [2] |
| \bar{A}, A^c | Complement set (given U) |
| $\#A, A $ | Cardinality |
| $a \in A$ | a is element of A |
| $a \notin A$ | a is not element of A |
| $\{1, 2, \dots, n\}$ | Discrete set containing the integer elements $1, 2, \dots, n$ |
| U | Universe |
| 2^A | Power set of A |
| \mathbb{R} | Set of real numbers |
| \mathbb{C} | Set of complex numbers |
| \mathbb{Z} | Set of integer number |
| $\mathbb{B} = \{0, 1\}$ | Boolean set |
| \emptyset | Empty set |
| \mathbb{N} | Set of natural numbers |
| $\mathbb{K} \in \{\mathbb{R}, \mathbb{C}\}$ | Real or complex space (field) |
| $\mathbb{K}^{I_1 \times I_2 \times \cdots \times I_N}$ | $I_1 \times I_2 \times \cdots \times I_N$ -dimensional real (or complex) space |
| \mathbb{K}_+ | Nonnegative real (or complex) space [2] |
| \mathbb{K}_{++} | Positive real (or complex) space, i.e., $\mathbb{K}_{++} = \mathbb{K}_+ \setminus \{\mathbf{0}\}$ [2] |
| $\mathbb{S}^n, \mathcal{S}^n$ | Conic set of the symmetric matrices in $\mathbb{R}^{n \times n}$ [2] |
| $\mathbb{S}_+^n, \mathcal{S}_+^n$ | Conic set of the symmetric positive semidefinite matrices in $\mathbb{R}^{n \times n}$ [2] |

| | |
|---|--|
| $\mathbb{S}_{++}^n, \mathcal{S}_{++}^n$ | Conic set of the symmetric positive definite matrices in $\mathbb{R}^{n \times n}$, i.e., $\mathbb{S}_{++}^n = \mathbb{S}_+^n \setminus \{\mathbf{0}\}$ [2] |
| \mathbb{H}^n | Set of all hermitian matrices in $\mathbb{C}^{n \times n}$ |
| $[a, b]$ | Closed interval of a real set from a to b |
| (a, b) | Opened interval of a real set from a to b |
| $[a, b), (a, b]$ | Half-opened intervals of a real set from a to b |

7.2 Quantifiers, inferences

| | |
|--------------|--|
| \forall | For all (universal quantifier) [8] |
| \exists | There exists (existential quantifier) [8] |
| \nexists | There does not exist [8] |
| $\exists!$ | There exist an unique [8] |
| \in | Belongs to [8] |
| \notin | Does not belong to [8] |
| \because | Because [8] |
| $, :$ | Such that, sometimes that parantheses is used [8] |
| $., (\cdot)$ | Used to separate the quantifier with restricted domain from the its scope, e.g., $\forall x < 0 (x^2 > 0)$ or $\forall x < 0, x^2 > 0$ [8] |
| \therefore | Therefore [8] |

7.3 Propositional Logic

| | |
|-----------------------|--|
| $\neg a$ | Logical negation of a [17] |
| $a \wedge b$ | Conjunction (logical AND) operator between a and b [17] |
| $a \vee b$ | Disjunction (logical OR) operator between a and b [17] |
| $a \oplus b$ | Exclusive OR (logical XOR) operator between a and b [17] |
| $a \rightarrow b$ | Implication (or conditional) statement [17] |
| $a \leftrightarrow b$ | Bi-implication (or biconditional) statement, i.e., $(a \rightarrow b) \wedge (b \rightarrow a)$ [17] |

$$a \equiv b, a \iff b, a \Leftrightarrow b$$

Logical equivalence, i.e., $a \leftrightarrow b$ is a tautology[17]

8 Physics

| | |
|--------------------------|---|
| E | Electric feild vector (in V/m) |
| Φ | Electric flux (scalar) (in V m) |
| D | Electric flux density, electric displacement, or electric induction vector (in C/m ²) |
| J | Electric current density vector (in A/m ²) |
| H | Magnetic feild vector (in A/m) |
| B | Magnetic flux density vector (in Wb/m ² = T) |
| ϵ | Electric permittivity |
| μ | Magnetic permeability |
| μ_0 | Magnetic permeability in vacuum |

9 Number theory, algorithm theory, and other notations

9.1 Mathematical symbols

| | |
|------------------|---------------------|
| ■ | Q.E.D. |
| \triangleq | Equal by definition |
| $:=, \leftarrow$ | Assignment [17] |
| \neq | Not equal |
| ∞ | Infinity |
| j | $\sqrt{-1}$ |

9.2 Calculus

| | |
|-----------|---|
| ∇ | Nabla operator (vector differential operator) |
| \oint_C | Closed line integral around the contour C |
| \iint_S | Sufarce integral over S enclosed by C |

9.3 Operations

| | |
|---------------------------|--|
| $ a $ | Absolute value of a |
| \log | Base-10 logarithm or decimal logarithm |
| \ln | Natural logarithm |
| $\operatorname{Re}\{x\}$ | Real part of x |
| $\operatorname{Im}\{x\}$ | Imaginary part of x |
| $\angle \cdot$ | phase (complex argument) |
| $x \bmod y$ | Remainder, i.e., $x - y\lfloor x/y \rfloor$, for $y \neq 0$ |
| $x \operatorname{div} y$ | Quotient [17] |
| $x \equiv y \pmod{m}$ | Congruent, i.e., $m \mid (x - y)$ [17] |
| $\operatorname{frac}(x)$ | Fractional part, i.e., $x \bmod 1$ [8] |
| $a \setminus b, a \mid b$ | b is a positive integer multiple of a , i.e., $\exists n \in \mathbb{Z}_{++} \mid b = na$ [8, 17] |
| $a \nmid b, a \nmid b$ | b is not a positive integer multiple of a , i.e., $\nexists n \in \mathbb{Z}_{++} \mid b = na$ [8, 17] |
| $\lceil \cdot \rceil$ | Ceiling operation [8] |
| $\lfloor \cdot \rfloor$ | Floor operation [8] |

9.4 Functions

| | |
|--|-----------------------|
| $\mathcal{O}(\cdot), \mathcal{O}(\cdot)$ | Big-O notation |
| $\Gamma(\cdot)$ | Gamma function |
| $\mathcal{Q}(\cdot)$ | Quantization function |

10 Abbreviations

| | |
|------|--|
| wrt. | With respect to |
| st. | Subject to |
| iff. | If and only if |
| EVD | Eigenvalue decomposition, or eigendecomposition [13] |
| SVD | Singular value decomposition |
| CP | CANDECOMP/PARAFAC |

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