### Notation

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### 1 Font notation

$a, b, c, \ldots, A, B, C, \ldots$	$\operatorname{Scalars}$
$a, b, c, \dots$	Vectors
$A, B, C, \dots$	Matrices
$\mathcal{A},\mathcal{B},\mathcal{C},\dots$	Tensors
$A, B, C, \ldots, A, B, C, \ldots, A, B, C, \ldots$	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 $n, k, m, i, \dots$ (parenthe-
$x_n, x_k, x_m, x_i, \dots$	sis should be adopted only if there
$x(n), x(k), x(m), x(i), \dots$	are no continuous-time signals in the
	context to avoid ambiguity)

#### 2.2 Common functions

$x_I(t)$ or $x_I[n]$	Real or in-phase part of $x(t)$ or $x[n]$
$x_Q(t)$ or $x_Q[n]$	Imaginary or quadrature part of $x(t)$
	or $x[n]$
$\delta(t)$	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

$f:A\to B$	A function $f$ whose domain is $A$ and
	codomain is $B$
$f^n, x^n(t), x^n[k]$	nth power of the function $f$ , $x[n]$ or
	x(t)
$f^{(n)}, x^{(n)}(t)$	nth derivative of the function $f$ or
	x(t)
$f', f^{(1)}, x'(t)$	1th derivative of the function $f$ or
J 7 J 7 1 (1)	x(t)
$f'', f^{(2)}, x''(t)$	2th derivative of the function $f$ or
$f^{\prime}$ , $f^{\prime}$ , $\chi^{\prime}$ (t)	
	x(t)
$\inf_{\mathbf{y} \in \mathcal{A}} g(\mathbf{x}, \mathbf{y})$	Infimum
	Cummorous
$\sup_{\mathbf{y} \in \mathcal{A}} g(\mathbf{x}, \mathbf{y})$	Supremum
$f \circ g$	Composition of the functions $f$ and
	g
*	Convolution
$\circledast, (N)$	Circular convolution
$x\left[\left((n-m)\right)_{N}\right],x\left((n-m)\right)_{N}$	Circular shift in $m$ samples within a
•	N-samples window

### 2.4 Transformations

$\mathcal{F}\left\{ \cdot  ight\}$	Fourier transform
$\mathcal{L}\left\{ \cdot  ight\}$	Laplace transform
$\mathcal{Z}\left\{ \cdot  ight\}$	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 fre-
	quency, 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 $x(t)$

$\tilde{X}[k]$ ,	$\tilde{X}(k)$ ,	$\tilde{X}_k$
X(z)		

Discrete Fourier series (DFS) of  $\tilde{x}[n]$  z-transform of x[n]

## 3 Probability, statistics, and stochastic processes

### 3.1 Operators and symbols

$E\left[\cdot\right]$	Statistical expectation
$E_u\left[\cdot\right]$	Statistical expectation with respect
	to u
$\mu_{\scriptscriptstyle X}$	Mean of the random variable $x$
$\mu_x, m_x$	Mean vector of the random variable
	X
$\mu_n$	nth-order moment of a random vari-
	able
$\mathcal{K}_x, \mu_4$	Kurtosis (4th-order moment) of the
	random variable $x$
$VAR[\cdot]$	Variance operator
$VAR_u[\cdot]$	Variance operator with respect to $u$
$\kappa_n$	nth-order cumulant of a random vari-
	able
$\sigma_x$ , $\kappa_2$	Variance of the random variable $x$
$\rho_{x,y}$	Pearson correlation coefficient be-
	tween $x$ and $y$
$a \sim P$	Random variable $a$ with distribution
	P

### 3.2 Stochastic processes

$r_{X}(\tau), R_{X}(\tau)$	Autocorrelation function of the sig-
	$\operatorname{nal} x(t) \text{ or } x[n]$
$S_x(f), S_x(j\omega)$	Power spectral density (PSD) of $x(t)$
	in linear $(f)$ or angular $(\omega)$ frequency
$S_{x,y}(f), S_{x,y}(j\omega)$	Cross PSD of $x(t)$ and $y(t)$ in linear
	or angular $(\omega)$ frequency
$\mathbf{R}_{\mathbf{x}}$	(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 $d(t)$
$\mathbf{R}_{\mathbf{x}\mathbf{y}}$	Cross-correlation matrix of $\mathbf{x}(n)$ and
•	$\mathbf{y}(n)$
$\mathbf{p}_{\mathbf{x}d}$	Cross-correlation vector between
	$\mathbf{x}(n)$ and $d(n)$

$$c_x(\tau), C_x(\tau)$$

 $C_x, K_x, \Sigma_x$   $c_{xy}(\tau), C_{xy}(\tau)$ 

 $C_{xy}, K_{xy}, \Sigma_{xy}$ 

# Autocovariance function of the signal x(t) or x[n]

(Auto)covariance matrix of **x** 

Cross-covariance function of the sig-

nal x(t) or x[n]

Cross-covariance matrix of  $\mathbf{x}$  and  $\mathbf{y}$ 

#### 3.3 Functions

 $Q(\cdot)$ 

 $\operatorname{erf}(\cdot)$ 

 $\operatorname{erfc}(\cdot)$ 

P[A]

 $p(\cdot), f(\cdot)$ 

 $p(x \mid A)$  $F(\cdot)$ 

 $\Phi_x(\omega), M_x(j\omega), E\left[e^{j\omega x}\right]$ 

 $M_x(t), \Phi_x(-jt), E[e^{tx}]$ 

 $\Psi_X(\omega), \ln \Phi_X(\omega), \ln E \left[e^{j\omega x}\right]$  $K_X(t), \ln E \left[e^{tx}\right], \ln M_X(t)$  *Q*-function, i.e.,  $P[\mathcal{N}(0,1) > x]$ 

Error function

Complementary error function i.e.,

 $\operatorname{erfc}(x) = 2Q(\sqrt{2}x) - \operatorname{erf}(x)$ 

Probability of the event or set A

Probability density function (PDF) or probability mass function (PMF)

Conditional PDF or PMF

Cumulative distribution function (CDF)

(CDF)

First characteristic function (CF) of

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Moment-generating function (MGF)

of x

Second characteristic function

Cumulant-generating function

(CGF) of x

#### 3.4 Distributions

 $\mathcal{N}(\mu, \sigma^2)$ 

 $\mathcal{CN}(\mu, \sigma^2)$ 

 $\mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$ 

 $\mathcal{CN}(\mu, \Sigma)$ 

 $\mathcal{U}(a,b)$  $\chi^2(n), \chi_n^2$  Gaussian distribution of a random variable with mean  $\mu$  and variance  $\sigma^2$  Complex Gaussian distribution of a random variable with mean  $\mu$  and variance  $\sigma^2$ 

Gaussian distribution of a vector random variable with mean  $\mu$  and covariance matrix  $\Sigma$ 

Complex Gaussian distribution of a vector random variable with mean  $\mu$  and covariance matrix  $\Sigma$ 

Uniform distribution from a to b Chi-square distribution with n degree of freedom (assuming that the Gaus-

sians are  $\mathcal{N}(0,1)$ 

$\operatorname{Exp}(\lambda)$	Exponential distribution with rate parameter $\lambda$
$\Gamma(lpha,eta)$	Gamma distribution with shape parameter $\alpha$ and rate parameter $\beta$
$\Gamma(lpha, heta)$	Gamma distribution with shape parameter $\alpha$ and scale parameter $\theta = 1/\beta$
$\operatorname{Nakagami}(m,\Omega)$	Nakagami-m distribution with shape parameter $m$ and spread parameter $\Omega$
$\operatorname{Rayleigh}(\sigma)$	Rayleigh distribution with scale parameter $\sigma$
$\operatorname{Rayleigh}(\Omega)$	Rayleigh distribution with the second moment $\Omega = E[x^2] = 2\sigma^2$
$\mathrm{Rice}(s,\sigma)$	Rice distribution with noncentrality parameter (specular component) $s$ and $\sigma$
$\mathrm{Rice}(A,K)$	Rice distribution with Rice factor $K=s^2/2\sigma^2$ and scale parameter $A=s^2+2\sigma^2$

## 4 Statistical signal processing

$\mathbf{\nabla} f, \mathbf{g}$	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 $x[n]$
$\hat{\boldsymbol{\mu}}_{x},\hat{\mathbf{m}}_{x}$	Sample mean of $x[n]$ or $x(t)$
$\hat{\boldsymbol{\mu}}_{\mathbf{x}},\hat{\mathbf{m}}_{\mathbf{x}}$	Sample mean vector of $\mathbf{x}[n]$ or $\mathbf{x}(t)$
$\hat{r}_{\scriptscriptstyle X}( au),\hat{R}_{\scriptscriptstyle X}( au)$	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
	$(\omega)$ frequency
$\hat{\mathbf{R}}_{\mathbf{x}}$	Sample (auto)correlation matrix
$\hat{r}_{x,d}( au), \hat{R}_{x,d}( au)$	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 $(\omega)$ frequency

$\hat{\mathbf{R}}_{\mathbf{x}\mathbf{y}}$	Sample cross-correlation matrix of
$\hat{ ho}_{x,y}$	$\mathbf{R}_{xy}$ Estimated Pearson correlation coefficient between $x$ and $y$
$\hat{c}_x( au), \hat{C}_x( au)$	Estimated autocovariance function of the signal $x(t)$ or $x[n]$
$\hat{\mathbf{C}}_{\mathbf{x}},\hat{\mathbf{K}}_{\mathbf{x}},\hat{\mathbf{\Sigma}}_{\mathbf{x}}$	Sample (auto)covariance matrix
$\hat{c}_{xy}(\tau), \hat{C}_{xy}(\tau)$	Estimated cross-covariance function
	of the signal $x(t)$ or $x[n]$
$\hat{ extbf{C}}_{ ext{xy}}, \hat{ extbf{K}}_{ ext{xy}}, \hat{ extbf{\Sigma}}_{ ext{xy}}$	Sample cross-covariance matrix
$\mathbf{w}, \mathbf{\theta}$	Parameters, coefficients, or weights vector
$\mathbf{w}_o, \mathbf{w}^{\star}, \mathbf{\theta}_o, \mathbf{\theta}^{\star}$	Optimum value of the parameters, coefficients, or weights vector
$\mathbf{W}$	Matrix of the weights
J	Jacobian matrix
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
P	Projection matrix; Permutation ma-
	trix
J	Jordan matrix
$\mathbf{L}$	Lower matrix
$\mathbf{U}$	Upper matrix
$\mathbf{C}$	Cofactor matrix
$\mathbf{C}_{\mathbf{A}}, \operatorname{cof}\left(\mathbf{A}\right)$	Cofactor matrix of <b>A</b>
S	Symmetric matrix
Q	Orthogonal matrix
$\mathbf{I}_N$	$N \times N$ -dimensional identity matrix
$0_{M imes N}$	$M \times N$ -dimensional null matrix
$0_N$	N-dimensional null vector
$1_{M imes N}$	$M \times N$ -dimensional ones matrix
$1_N$	N-dimensional ones vector
0	Null matrix, vector, or tensor (di-
	mensionality understood by context)
1	Ones matrix, vector, or tensor (di-
	mensionality understood by context)

### 5.2 Indexing

$x_{i_1,i_2,\ldots,i_N}, [\boldsymbol{\mathcal{X}}]_{i_1,i_2,\ldots,i_N}$	Element in the position $(i_1, i_2, \ldots, i_N)$ of the tensor $\mathcal{X}$
$\mathcal{X}^{(n)}$	nth tensor of a nontemporal sequence
$\mathbf{x}_n, \mathbf{x}_{:n}$	nth column of the matrix $X$
$\mathbf{x}_{n}$ :	nth row of the matrix $X$
$\mathbf{X}_{i_1,,i_{n-1},:,i_{n+1},,i_N}$	Mode- $n$ fiber of the tensor $\boldsymbol{\mathcal{X}}$
$\mathbf{X}_{:,i_2,i_3}$	Column fiber (mode-1 fiber) of the
	thrid-order tensor $\boldsymbol{\mathcal{X}}$
$\mathbf{x}_{i_1,:,i_3}$	Row fiber (mode-2 fiber) of the thrid-
	order tensor $\boldsymbol{\mathcal{X}}$
$\mathbf{x}_{i_1,i_2,:}$	Tube fiber (mode-3 fiber) of the
	thrid-order tensor $\boldsymbol{\mathcal{X}}$
$\mathbf{X}_{i_1,:,:}$	Horizontal slice of the thrid-order
	tensor $\boldsymbol{\mathcal{X}}$
$\mathbf{X}_{:,i_2,:}$	Lateral slices slice of the thrid-order
, 2,	tensor $\boldsymbol{\mathcal{X}}$
$\mathbf{X}_{i_3},\mathbf{X}_{:,:,i_3}$	Frontal slices slice of the thrid-order
.0 77.0	tensor $\mathcal{X}$

### 5.3 Operations with tensors

 $\mathbf{X}_{(n)}$   $n ext{-mode matricization of the tensor } \boldsymbol{\mathcal{X}}$ 

### 5.4 Operations with matrices

$\mathbf{A}^{-1}$	Inverse matrix
$\mathbf{A}^+,\mathbf{A}^\dagger$	Moore-Penrose pseudoinverse
$\mathbf{A}^{\top}$	Transpose
$\mathbf{A}^{- op}$	Transpose of the inverse
$\mathbf{A}^*$	Complex conjugate
$\mathbf{A}^H$	Hermitian
$\ \mathbf{A}\ _{\mathrm{F}}$	Frobenius norm
$\ \mathbf{A}\ $	Matrix norm
$ \mathbf{A} , \det{(\mathbf{A})}$	Determinant
$\operatorname{diag}\left(\mathbf{A}\right)$	The elements in the diagonal of ${\bf A}$
$\text{vec}(\mathbf{A})$	Vectorization: stacks the columns of
	the matrix A into a long column vec-
	tor
$\operatorname{vec}_{\operatorname{d}}\left(\mathbf{A}\right)$	Extracts the diagonal elements of a square matrix and returns them in a
	column vector

$\operatorname{vec}_{\operatorname{l}}\left(\mathbf{A}\right)$	Extracts the elements strictly below
	the main diagonal of a square matrix
	in a column-wise manner and returns
	them into a column vector
$\operatorname{vec}_{\mathrm{u}}\left(\mathbf{A}\right)$	Extracts the elements strictly above
	the main diagonal of a square matrix
	in a column-wise manner and returns
	them into a column vector
$\operatorname{vec_b}\left(\mathbf{A}\right)$	Block vectorization operator: stacks
	square block matrices of the input
	into a long block column matrix
$\operatorname{unvec}\left(\mathbf{A}\right)$	Reshapes a column vector into a ma-
	trix
$\mathrm{tr}\left(\mathbf{A} ight)$	trace
$\otimes$	Kronecker product
$\odot$	Hadamard (or Schur) (elementwise)
	product
$\mathbf{A}^{\odot n}$	nth-order Hadamard power of the
	matrix <b>A</b>
$\mathbf{A}^{\odot \frac{1}{n}}$	nth-order Hadamard root of the ma-
	trix A
Ø	Hadamard (or Schur) (elementwise)
	division
<b>♦</b>	Khatri-Rao product
⊗	Kronecker Product
$\times_n$	<i>n</i> -mode product
$^{n}$	" mode product

### 5.5 Operations with vectors

$\ \mathbf{a}\ $	$l_1$ norm, 1-norm, or Manhatan 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
$\ \mathbf{a}\ _{\infty}^{\cdot}$	$l_{\infty}$ norm, $\infty$ -norm, or Chebyshev
	norm
$\operatorname{diag}\left(\mathbf{a}\right)$	Diagonalization: a square, diagonal
	matrix with entries given by the vec-
	tor <b>a</b>
$\langle \mathbf{a}, \mathbf{b}  angle$	Inner product, i.e., $\mathbf{a}^{T}\mathbf{b}$
$\mathbf{a} \circ \mathbf{b}$	Outer product, i.e., $\mathbf{a}\mathbf{b}^{T}$

### 5.6 Decompositions

Λ

Eigenvalue matrix

Q	Eigenvectors matrix; Orthogonal ma-
R	trix of the QR decomposition Upper triangular matrix of the QR decomposition
U	Left singular vectors
$\mathbf{U}_r$	Left singular nondegenerated vectors
$\Sigma$	Singular value matrix
$\Sigma_r$	Singular value matrix with nonzero
	singular values in the main diagonal
$\Sigma^+$	Singular value matrix of the pseu-
	doinverse
$\Sigma_r^+$	Singular value matrix of the pseu-
	doinverse with nonzero singular val-
	ues in the main diagonal
V	Right singular vectors
$\mathbf{V}_r$	Right singular nondegenerated vec-
• (4)	tors
$\operatorname{eig}\left(\mathbf{A}\right)$	Set of the eigenvalues of <b>A</b>
$[\![\mathbf{A},\mathbf{B},\mathbf{C},\ldots]\!]$	CANDECOMP/PARAFAC (CP) de-
	composition of the tensor $\mathcal{X}$ from the
	outer product of column vectors of <b>A</b> ,
$[\![\boldsymbol{\lambda};\mathbf{A},\mathbf{B},\mathbf{C},\ldots]\!]$	B, C, Normalized CANDE-
$[\mathbf{A}, \mathbf{A}, \mathbf{D}, \mathbf{C}, \ldots]$	CANDE- COMP/PARAFAC (CP) decom-
	position of the tensor $\mathcal{X}$ from the
	outer product of column vectors of
	A, B, C,
	<u> </u>

## 5.7 Spaces

$N(\mathbf{A})$ , nullspace( $\mathbf{A}$ ), kernel( $\mathbf{A}$ ) $C(\mathbf{A})$ , columnspace( $\mathbf{A}$ ), range( $\mathbf{A}$ )	Nullspace (or kernel space) Columnspace (or range), i.e., the space span $(\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n)$ , where $\mathbf{a}_i$ is the ith column vector of the matrix
	A
$\mathrm{span}\left(\mathbf{a}_{1},\mathbf{a}_{2},\ldots,\mathbf{a}_{n}\right)$	Vector space spanned by the argument vectors
$\mathrm{span}\left(\mathbf{A}\right)$	Vector space spanned by the col- umn vectors of <b>A</b> , which gives the columnspace of <b>A</b>
$\operatorname{rank}\left(\mathbf{A} ight)$	Rank, that is, $\dim (\operatorname{span} (\mathbf{A})) = \dim (\mathbf{C} (\mathbf{A}))$
nullity (A)	Nullity of $\mathbf{A}$ , i.e., dim $(N(\mathbf{A}))$
$\mathbf{a} \perp \mathbf{b}$	$\mathbf{a}$ is orthogonal to $\mathbf{b}$
a ≠ b	${f a}$ is not orthogonal to ${f b}$

#### 5.8 Inequalities

$\mathcal{X}$	$\leq$	0
a :	≤ <sub>K</sub>	b

 $\mathbf{a} \prec_K \mathbf{b}$ 

 $\mathbf{a} \leq \mathbf{b}$ 

 $\mathbf{a} \prec \mathbf{b}$ 

 $\mathbf{A} \leq_K \mathbf{B}$ 

 $\mathbf{A} \prec_K \mathbf{B}$ 

 $A \leq B$ 

A < B

Nonnegative tensor

Generalized inequality meaning that  $\mathbf{b} - \mathbf{a}$  belongs to the conic subset K in the space  $\mathbb{R}^n$ 

Strict generalized inequality meaning that  $\mathbf{b} - \mathbf{a}$  belongs to the interior of the conic subset K in the space  $\mathbb{R}^n$  Generalized inequality meaning that  $\mathbf{b} - \mathbf{a}$  belongs to the nonnegative orthant conic subset,  $\mathbb{R}^n_+$ , in the space  $\mathbb{R}^n$ .

Strict generalized inequality meaning that  $\mathbf{b} - \mathbf{a}$  belongs to the positive orthant conic subset,  $\mathbb{R}^n_{++}$ , in the space  $\mathbb{R}^n$ 

Generalized inequality meaning that  $\mathbf{B} - \mathbf{A}$  belongs to the conic subset K in the space  $\mathbb{S}^n$ 

Strict generalized inequality meaning that  $\mathbf{B} - \mathbf{A}$  belongs to the interior of the conic subset K in the space  $\mathbb{S}^n$  Generalized inequality meaning that  $\mathbf{B} - \mathbf{A}$  belongs to the positive semidefinite conic subset,  $\mathbb{S}^n_+$ , in the space  $\mathbb{S}^n$  Strict generalized inequality meaning that  $\mathbf{B} - \mathbf{A}$  belongs to the positive orthant conic subset,  $\mathbb{S}^n_{++}$ , in the space  $\mathbb{S}^n$ 

#### 6 Sets

A + B
A - B
$A \setminus B, A - B$

 $A \cup B$   $A \cap B$   $A \times B$   $A^n$ 

Set addition (Minkowski sum)

Minkowski difference

Set difference or set subtraction, i.e., the set containing the elements

of A that are not in B

Set of union Set of intersection Cartesian product  $A \times A \times \cdots \times A$ 

n times

$A^{\perp}$	Orthogonal complement of $A$ , e.g., $N(\mathbf{A}) = C(\mathbf{A}^{\top})^{\perp}$
$A \oplus B$	Direct sum, e.g., $C(\mathbf{A}^{\top}) \oplus C(\mathbf{A}^{\top})^{\perp} = \mathbb{R}^{n}$
$A^c,ar{A}$	Complement set (given $U$ )
#A,  A	Cardinality
$a \in A$	a is element of $A$
$a \notin A$	a is not element of $A$
$\{1,2,\ldots,n\}$	Discrete set containing the integer el-
	ements $1, 2, \ldots, n$
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
Ø	Empty set
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
$\mathbb{K}_{++}$	Positive real (or complex) space, i.e.,
	$\mathbb{K}_{++} = \mathbb{K}_{+} \setminus \{0\}$
$\mathbb{S}^n, \mathcal{S}^n$	Conic set of the symmetric matrices
	in $\mathbb{R}^{n \times n}$
$\mathbb{S}^n_+, \mathcal{S}^n_+$	Conic set of the symmetric positive
	semidefinite matrices in $\mathbb{R}^{n \times n}$
$\mathbb{S}^n_{++}, \mathcal{S}^n_{++}$	Conic set of the symmetric positive
	definite matrices in $\mathbb{R}^{n\times n}$ , i.e., $\mathbb{S}^n_{++}$ =
	$\mathbb{S}^n_+\setminus\{0\}$
$\mathbb{H}^n$	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 Communication systems

Low-pass equivalent signal or enve $s_l$ lope complex of sGaussian noise  $\eta, w$ Received signal r τ Timming delay Timming error (delay - estimated)  $\Delta \tau$ Phase offset  $\varphi$ Phase error (offset - estimated)  $\Delta \varphi$ Doppler frequency  $f_d$ Received signal amplitude  $\boldsymbol{A}$ Combined effect of the path loss and γ antenna gain

#### 8 Other notations

#### 8.1 Mathematical symbols

Ξ There exists ∄ There does not exist ∃! There exist an unique  $\in$ Belongs to ∉ Does not belong to Q.E.D. Therefore Because  $\forall$ For all **|**,: Such that Logical equivalence Equal by definition ≜,:= Not equal Infinity  $\infty$  $\sqrt{-1}$ j

#### 8.2 Operations

 $\mathrm{Im}\,\{x\}$ Imaginary part of x۷٠ phase (complex argument)  $x \mod y$ Remainder, i.e.,  $x - y \lfloor x/y \rfloor$  $\operatorname{frac}(x)$ Fractional part, i.e.,  $x \mod 1$  $a \wedge b$ Logical AND of a and b $a \vee b$ Logical OR of a and bLogical negation of a $\neg a$  $\lceil \cdot \rceil$ Ceiling operation Floor operation  $\lfloor \cdot \rfloor$ 

#### 8.3 Functions

 $\begin{array}{ll} \mathcal{O}(\cdot), O(\cdot) & \text{Big-O notation} \\ \Gamma(\cdot) & \text{Gamma function} \end{array}$ 

#### 9 Abbreviations

wrt. With respect to st. Subject to iff. If and only if EVD Eigenvalue decomposition, or eigendecomposition

SVD Singular value decomposition CP CANDECOMP/PARAFAC