

# notation

## Notation

Symbol	Meaning
$c_1, c_2, \dots, c_z$	the thresholds for simplification
$C_0, C_1, \dots, C_z$	the groups for simplification
$f_\phi(\cdot)$	the function on temporal processing with parameters $\phi$
$F_\eta(\cdot)$	the PDF to fit the data with parameter $\eta$
$g_\theta(\cdot)$	the function on spatial aggregation with parameter $\theta$
$h(\cdot)$	the function on variable transformation, no parameter
$i : 1, 2, \dots, n$	the generic space index in the data
$j : 1, 2, \dots, J$	the generic time index in the data
$j^*$	variation of time index in normalising
$j'$	variation of time index in temporal processing
$k$	time scale in temporal processing
$m : 0, 1, \dots, 11$	remainder of $j^* \bmod 12$ for index monthly data
$n$	the maximum space index in the data
$p : 1, 2, \dots, P$	the (maximum) variable index in the data
$p^* : 1, 2, \dots, P^*$	the dimension reduced variable index
$q(\cdot)$	the function on dimension reduction
$u$	the subtracted constant in benchmarking
$\mathcal{D}_s$ and $\mathcal{D}_t$	the parameter space for space and time
$\alpha$	the subtracted quantity in scaling
$\gamma$	the divided quantity in scaling
$\eta$	parameters in PDF distribution fit
$\theta$	additional parameters in spatial aggregation
$\lambda$	the PCA factor loadings
$\sigma$	the standard deviation of the data in scaling
$\phi$	additional parameters in temporal processing
$\Phi^{-1}(\cdot)$	the quantile (inverse CDF) function

Step	Notation	Notes
Raw data	$\mathbf{X}(\mathbf{s}; \mathbf{t}),$ $x_p(s_i; t_j)$	$\mathbf{s} \in \mathcal{D}_s, \mathcal{D}_s \subseteq \mathbb{R}^2, \mathbf{s} = (s_1, s_2, \dots, s_n)'$ $t \in \mathcal{D}_t, \mathbf{t} = (t_1, t_2, \dots, t_J)'$ $\mathbf{X}(\mathbf{s}; \mathbf{t}) = (x_1(\mathbf{s}; \mathbf{t}), x_2(\mathbf{s}; \mathbf{t}), \dots, x_P(\mathbf{s}; \mathbf{t}))'$ when the pipeline step can be written in univariate case, the data will be referred to as $x(\mathbf{s}; \mathbf{t})$
Spatial aggregation	$g_\theta(x(\mathbf{s}; \mathbf{t}))$	TODO: to be filled
scaling	$\frac{x(s_i; t_j) - \alpha}{\gamma}$	Example 1: z-score standardisation: $\alpha = \bar{x}(s; t)$ and $\gamma = \sigma(s; t)$ to standardise the data by the its mean and standard deviation across all time and all space units. Example 2: standardise into unit interval: $\alpha = \min[x(s_i, t_j)]$ and $\gamma = \max[x(s_i, t_j)] - \min[x(s_i, t_j)]$ where $\min[\cdot]$ and $\max[\cdot]$ are the minimum and maximum value across all spatial and temporal units.
Normalising	$\Phi^{-1}[F_\eta(x(\mathbf{s}; \mathbf{t}))]$	Example: When PDF $F(\cdot)$ is separately fitted for each month: $\Phi^{-1}[F_\eta^m(x(\mathbf{s}; t_{j^*}))]$ where $j^*$ is all the indexes that satisfy $j^* \bmod 12 = m$ for each $m = 0, 1, \dots, 11$
Variable transformation	$h(x(\mathbf{s}; \mathbf{t}))$	TODO: to be filled
Temporal processing	$f_\phi(x(\mathbf{s}; \mathbf{t}))$	Example 1: aggregate across a time scale of $\phi = k$ : $x(s_i; t_{j'}) = \sum_{j=j'-k+1}^{j'} x(s_i; t_j)$ Example 2: aggregate with a kernel weight $\phi = w_{ij}$ : $x(s_i; t_{j'}) = \sum_{j=j'-k+1}^{j'} w_{ij} x(s_i; t_j)$
Dimension reduction	$x_{p^*}(\mathbf{s}; \mathbf{t}) = q(x_p(\mathbf{s}; \mathbf{t}))$ where $p^* = 1, 2, \dots, P^*$ and $P^* < P$	Example: first principal component: $x_{p^*}(\mathbf{s}; \mathbf{t}) = \sum_{p=1}^P \lambda_p x_p(\mathbf{s}; \mathbf{t})$ where $\lambda_p$ is the loading of the PC1, derived from maximising the variance of the data given the constraint $\sum_{p=1}^P \lambda_p^2 = 1$
benchmarking	$\ln(s_i, t_j) - u$	where $u$ is a constant of interest in the index constructed. Example: In SPI, $u = -2$ is the threshold for extreme drought.

Step	Notation	Notes
Simplification	$C_0 \quad c_1 \leq (s_i; t_j) < c_0$	<u>Example:</u> In SPI, four categories are classified: mild drought: $[-0.99, 0]$ ; moderate drought: $[-1.49, -1]$ ; severe drought: $[-1.99, -1.5]$ , and extreme drought: $[-\infty, -2]$ . Here $C_0, C_1, C_2, C_3$ are the drought categories: mild, moderate, severe, and extreme drought ( $z = 3$ ) and $c_0 = 0, c_1 = -1, c_2 = -1.5, c_3 = -2$
	$C_1 \quad c_2 \leq x(s_i; t_j) < c_1$	
	$C_2 \quad c_3 \leq x(s_i; t_j) < c_2$	
	...	
	$C_z \quad c_z \leq x(s_i; t_j)$	