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## NOMENCLATURE

a	dimensionless scale factor
A, B, C	dimensionless elements of $2^{\Theta}$
b	time shift, s
bet	pignistic probability
Cr	credibility
$d_{n_{node},n_{level}}(k)$	$k^{th}$ coefficient of a node at $n_{node}, n_{level}$
E	energy, J
Fmeasure	dimensionless weighted harmonic mean of
	precision and recall
J	dimensionless constant
k	dimensionless conflict factor
m	dimensionless belief mass
Nai	dimensionless number of vectors assigned
	to the class i
Ncai	dimensionless number of vectors correctly
	assigned to the class i
Ni	dimensionless number of vectors of the
	class i
n <sub>level</sub>	dimensionless level of wavelet
	decomposition
n <sub>node</sub>	dimensionless node position at the
	<i>n<sub>level</sub></i> <sup>th</sup> level
n <sub>coef</sub>	dimensionless number of coefficients of the
)	$n_{level}$ <sup>th</sup> level
s(t)	finite energy signal, volt
Pl	plausibility
S	source
S <sub>EMG</sub>	threshold of EMG signal, volt
t	time, s
$W_S$	wavelet transform applied to a finite energy
	signal $s(t)$
x	observation

## **Greek symbols**

- μ mean of envelope, volt
- $\sigma$  standard deviation of envelope, volt
- φ empty set
- $\psi$  mother wavelet
- Θ discernment framework
- $\theta_i$  Singleton