

$c_h [q, h]$	Artificial output function harmonic coefficient of dimensions q and h
$c_p [q, p]$	Artificial output function polynomial coefficient of dimensions q and p
$f (t [q]) \forall q \in [1, Q]$	Artificial output parameter
H	Maximum degree of harmonic fitting
h	Artificial output function harmonic coefficient degree dimension index
i	Index of measurement instance
P	Maximum degree of polynomial fitting
p	Artificial output function polynomial coefficient degree dimension index
Q	Input parameter dimension count
q	Input parameter dimensional index
$t [q]$	Artificial input parameter of dimension q
$x [q, i]$	Measured input parameter of dimension q at instance i
$y [i]$	Measured output parameter at instance i
$\delta [q, h]$	Phase offset of harmonic fitting of dimensions q and h
$\varepsilon [i]$	Phase offset of harmonic fitting of dimensions q and n
$\omega [q]$	Fundamental frequency of harmonic fitting of dimension q

This document describes the formulation of an algorithm to select parameters for a function that is fitted to a measured data set. The family of functions that are generated by this algorithm are a linear combination of polynomial and harmonic terms. This may be considered as a hybrid of Taylor-series and Fourier-series fitting.

Assume we are given a set of measurements. We arbitrarily select one dimension of these measurements as our output parameter set, $y [i]$. The remainder are the input parameter set, $x [q, i]$.

The goal of the algorithm is to determine the coefficients of the function that best match the measure dataset.

The function has the form described in equation (1).

$$f (t [q]) = c_p [1, 0] + \sum_{q=1}^{q=Q} \left(\left(\sum_{n=1}^{p=P} c_p [q, n] \cdot (t [q])^n \right) + \left(\sum_{h=1}^{h=H} c_h [q, h] \cdot \cos (h \cdot \omega [q] \cdot t [q] + \delta [q, h]) \right) \right) \quad (1)$$

$c_p [1, 0]$ is the static offset of the fitting function.

The error between the fitting function and the measured data is defined as in equation (2).

$$\varepsilon [i] = f (x [q, i]) - y [i] \quad (2)$$

The algorithm minimizes the net square error.

$$0 = \frac{\partial}{\partial c_h [q, h]} (\varepsilon^2 [i]) = \varepsilon [i] \cdot \frac{\partial \varepsilon [i]}{\partial c_h [q, h]} \quad (3a)$$

$$0 = \frac{\partial}{\partial \delta [q, h]} (\varepsilon^2 [i]) = \varepsilon [i] \cdot \frac{\partial \varepsilon [i]}{\partial \delta [q, h]} \quad (3b)$$

$$0 = \frac{\partial}{\partial \omega [q]} (\varepsilon^2 [i]) = \varepsilon [i] \cdot \frac{\partial \varepsilon [i]}{\partial \omega [q]} \quad (3c)$$

$$0 = \frac{\partial}{\partial c_p [q, p]} (\varepsilon^2 [i]) = \varepsilon [i] \cdot \frac{\partial \varepsilon [i]}{\partial c_p [q, p]} \quad (3d)$$

Each of the partial derivatives in equation (3) are expanded in equation (4)

$$\frac{\partial \varepsilon [i]}{\partial c_h [q, h]} = \cos (h \cdot \omega [q] x [q, i] + \delta [q, h]) \quad (4a)$$