# 962. On-board diagnostics of mechanical defects of the vehicle drive system based on the vibration signal reference model

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**Abstract.** The parametric model proposition of modelling vibrations signal of the vehicle drive system is presented in the paper. The reference model is established for the new drive system to which the control measurements, performed at definite time intervals of automobile running, are compared. This base model must be actualized after each drive system overhaul and as the casual wear progresses. The recursive identification method of the model is performed. The method allows the on-board diagnostics of a gearbox, main gearbox, differential gear and triple joints. The diagnostic method was verified on the example of the vibrations signal recorded for the drive system of the Fiat Punto automobile.

**Keywords:** vehicle drive system, reference model, vibration signal, gearbox vibrations, on-board diagnostics, mechanical defects.

# 1. Introduction

The model-based approach to diagnostics problem is proposed in the paper [1]. During the modelling process of the drive system the problem of selecting the complexity degree of the model occurs. Of course, it would be possible to design one global model of the vehicle drive system [2] and to expand it, but in such case the complication degree will not allow to use it for the on-board diagnostics. Instead of the structural model, which is too complicated, the statistic model of the vibroacoustic signal propagation – on a very high abstraction level – is proposed in the presented hereby study. The necessary condition of the proper model operation is then its accurate formal identification [3]. This model is created in the adaptive way, which means firstly: the accelerated model identification and secondly: adjusting the model to the given type of the drive system and its working time [4]. Designing of such a model requires performing a series of measurements in the selected points of the drive system as well as other additional synchronizing signals during normal vehicle maintenance, it means during driving at a steady speed and without abrupt load changes. The base model takes into account the influence of responses of all elements of the drive system. The proposed method of modelling can be used for the on-board diagnostics of the mechanical defects of the vehicle drive line.

#### 2. Vibration signal measurement

Signals of approximately 1 second duration are recorded while driving at constant rotational speed, without abrupt changes of an engine load. Small fluctuations of rotational speeds are eliminated during further analysis by means of time synchronization and interpolation operations. Maintaining a constant rotational speed of the engine is essential, since this parameter strongly influences the vibration amplitude.

The following signals are needed to perform the described method:

- accelerations of the gearbox housing vibrations;
- voltage from the sensor of the crankshaft position (for a speed estimation);
- voltage from the sensor of the throttle position (for a load estimation).

Accelerations of vibration signals of the drive system, due to the narrower frequency range of the investigated signal, were processed by means of the Bruel & Kjaer sensors of type IEPE No. 4514. Signals were recorded by means of the portable recording device Bruel & Kjaer PULSE type 3560E with the sampling frequency of 65536 Hz. The crankshaft position signal and the coil signal allowed the identification of engine work cycles, ignition moments and timing gear phases.

Signals of vibration accelerations of the vehicle drive system are periodical, and for the constant rotational speed also stationary and ergodic with elements of a random noise. The spectrum of the vibration signal generated by the gearbox is presented in Fig. 1. Its main components are trains of harmonics of the crank shaft rotational speed [5-7]. This is the broadband spectrum and the generation of so many harmonics is related, among others, to the unbalancing and misalignments of shafts – changing during the drive – as well as to transferring of vibrations caused by the cyclic operations of engine valves. The successive harmonics (for the 4-cylinder engine this is the half of the crank shaft frequency). The next excitation source constitutes the tooth-passing of the gearbox and the main gearbox and also their harmonics. In the range of low frequencies (up to the crank shaft rotational frequency  $f_0$ ) the component of the half-shaft and the components related to operations of individual cylinders are seen.



Fig. 1. Average spectrum of gearbox vibrations acceleration (vibration amplitude vs harmonic number of the crankshaft frequency) [4]: a) entire measurement range, b) low frequency range

## 3. The reference model and its identification

During the life time of the drive system its vibration characteristics are changing, spectrum

becomes smeared, new components are added, etc. To be the universal model it has to be able to adapt itself for various drive systems being in various states of wear. First of all, the reference model remembered for the new system must exist. Such model should automatically update itself after each repair or each change of parts. Finally, it must be determined for the selected engine rotational speeds since the vibration characteristics are strongly dependent on this parameter.

Consider the autoregressive (AR) signal model represented in the transfer function form:

$$A(z^{-1})y(k) = e(k),$$
(1)

where the polynomial  $A(z^{-1}) = 1 + a_1 z^{-1} + a_1 z^{-2} + ... + a_1 z^{-n}$ ,  $z^{-1}$  is the backward shift operator, y(k) – measurement, e(k) – zero mean white noise and  $a_i$ , i = 1, 2, ..., n are the unknown constant parameters for each sampling moment k.

The above AR model can be written in the form of linear regression relationship [8]:

$$y(k) = x^{\mathrm{T}}(k)a + e(k), \tag{2}$$

where:

$$x^{\mathrm{T}}(k) = [-y(k-1), -y(k-2), \dots, -y(k-n)],$$
(3)

and:

$$\boldsymbol{a} = [\boldsymbol{a}_1, \boldsymbol{a}_2, \dots \boldsymbol{a}_n]^{\mathrm{T}}.$$
(4)

Then the deterministic recursive least squares (RLS) estimation algorithm takes the following form:

1. 
$$\hat{a}(k) = \hat{a}(k-1) + g(k)E(k),$$
 (5)

2. 
$$g(k) = P(k-1)x(k)[1+x^{T}(k)P(k-1)x(k)]^{-1}$$
, (6)  
3.  $P(k) = P(k-1) - g(k)x^{T}(k)P(k-1)$ , (7)

3. 
$$P(k) = P(k-1) - g(k)x^{\mathrm{T}}(k)P(k-1),$$
 (7)

with the:

$$E(k) = y(k) - x^{\mathrm{T}}(k)\hat{a}(k-1),$$
(8)

representing the recursive residual or one-step-ahead prediction error for each sampling moment k, calculated on the basis of the actual measurements y(k) and the estimate  $\hat{a}(k-1)$ .

The Auto Regressive (AR) model is characterized by a lack of the input signal. In this case, the vibration response is treated as the result of the influence of a certain non-measurable disturbance e(k).

Before the identification the signal must be properly prepared by:

- removal of constant components and trends,
- filtration in order to select the characteristic frequency range,
- determination of the autocorrelation function or synchronous averaging,
- signal decimation.

Thus let us perform the parametric identification for the time-waveform of the gearbox vibration signal in the low-frequency range, of the drive system in good technical condition. This signal should be prepared by filtering out high frequencies (higher than three times the crankshaft frequency). Then the autocorrelation function must be determined. 10-times decimation was also applied since the signal was over-sampled and the model identification would require too many rather useless coefficients. Recursive identification of the model

parameters was performed.

On the basis of the model assessment indices, it can be assumed that the auto-regression model of the 4th order is satisfactory. The comparison of the recorded vibration signal with the AR model output of 4 coefficients a is presented in Fig. 2.



**Fig. 2.** Reference model of the gearbox low-frequency vibration signal: a) comparison of the waveform of the signal autocorrelation function and AR model, b) coefficients *a* of the AR model adapting during the recursive model identification

The model quality assessment is done on the basis of the following indices: Akaike's Information Criterion (AIC), Final Prediction Error (FPE), Fit [4, 9].

# 4. Case study of the gearbox wear diagnostics

Generating the reference spectrum and its comparison with the signal spectrum is the easiest method, however less suitable at on-board diagnostics of vehicle drive system defects [10]. If, in addition, we do not have the recorded waveforms long enough to obtain the sufficient resolution of the spectrum, the time waveforms seem more suitable.

Another reason of using the time waveform is a signal modulation, as in the case of uneven wear of gearbox teeth (Fig. 3).



Fig. 3. Worn teeth of the gearbox

The waveform of the vibrations acceleration signal for such a case is presented in Fig. 4. That time, the averaged vibrations spectrum is not as useful as the time waveform.

On the grounds of the filtered and synchronously averaged waveforms the basic measures of the vibration signal are determined. The selected measures of the signal determined for the gearbox in good technical condition and for the gearbox with worn teeth are listed in Table 1.

The list of measures can be useful for the subsequent confirmation of the diagnosis, but the main procedure is based on the diagnostic model.



**Fig. 4.** Synchronously averaged waveforms of the gearbox vibration at a speed of 3000 rpm for: the gearbox in good condition, the gearbox with casual wear of the teeth and the gearbox just before a failure

Tuble 1. Signal measures of the teeth wear of the gearbox at 5000 ipin					
Vibration signal measure	Gearbox in good technical condition	Casual wear	Before a failure		
Peak-to-peak value [m/s <sup>2</sup> ]	10.3008	33.6274	546.0568		
RMS value [m/s <sup>2</sup> ]	1.8343	4.2969	78.7564		
Average value [m/s <sup>2</sup> ]	1.4510	3.1217	51.3733		
Peak-to-peak value of the envelope [m/s <sup>2</sup> ]	5.8500	17.9173	284.4201		

Table 1. Signal measures of the teeth wear of the gearbox at 3000 rpm

The diagnostic system records, every determined amount of kilometres, the gearbox vibrations acceleration signal together with auxiliary signals mentioned above. Then it is subjected to further analysis, that is to filtration, decimation and determination of the autocorrelation function (as at the reference model determination). Such signal preparation eliminates random errors and alarms. This requires using the determined amount of the system memory and time, thus the procedure is not exactly on-line. The transformed signal of the vibrations acceleration of the worn pair of wheels of the V-th gear was compared in Fig. 5 with the reference model, for the characteristic frequency band.



**Fig. 5.** Comparison of the autocorrelation function of the signal for the reference model, the gearbox with casual wear of the teeth and the gearbox just before a failure

Using of the autocorrelation function guarantees removal of the uncorrelated noise and allows building of simpler model with small amount of coefficients. The not normalized autocorrelation function makes possible a direct comparison of the signals in Fig. 5 and can be used for the assessment of the level of wear. If the normalized autocorrelation function is used, only classification 'good – bad' is possible.

On the grounds of the adaptive reference model for the system in good technical condition the residual signals for the gearbox in good condition and for the worn one were determined (Fig. 6).



Fig. 6. Residual signals for three levels of wear of the teeth of gearbox

The selected measures of the residual signals determined for the gearbox in good technical condition and for the worn out gearbox are listed in Table 2.

<b>Tuble 1.</b> Residual measures for three fevers of the feeth wear of the genrook				
Residual measure	Gearbox in good condition	Casual wear Before a failure		
Peak-to-peak value [m/s <sup>2</sup> ]	0.1780	0.9549	423.2935	
RMS value [m/s <sup>2</sup> ]	0.0331	0.1657	58.4411	
Average value [m/s <sup>2</sup> ]	0.0257	0.1194	36.0910	
Uncertainty of the model	$1.9086 \cdot 10^{-4}$			

**Table 2.** Residual measures for three levels of the teeth wear of the gearbox

The uncertainty of the model was calculated according to the following formula:

$$u(x) = \sqrt{S_{\bar{x}}^2} = \sqrt{\frac{1}{n(n-1)} \sum_{i=1}^n (x_i - \bar{x})^2} = \sqrt{\frac{1}{n(n-1)} \sum_{i=1}^n (r_i)^2},$$
(9)

where  $S_{\bar{x}}$  – standard deviation;  $x_i$  – signal for gearbox with no defects (properly pre-processed);  $\bar{x}$  – model output;  $r_i$  – residual signal for gearbox with no defects; n – number of samples.

For confirmation of the diagnosis further signal analysis like spectrum of the signals and residuals should be checked.

## 5. Case study of the 'low-frequency' defects diagnostics

The methodology of proceedings can be presented for the worn differential gear (Fig. 7) or for the worn triple joint (Fig. 8).

The parametric identification for the time-waveform of the gearbox vibration signal in the low-frequency range, of the drive system in good technical condition was shown in Fig. 2a. In

this case the signal should be prepared by filtering out high frequencies (higher than three times the crankshaft frequency) and by synchronous averaging instead of the autocorrelation function. Such solution needs more complicated model (more coefficients of the model).

The comparison of the averaged signals of the base model low frequency vibration signal of the gearbox, with the signals recorded for the worn differential gear and for the worn triple joint, is presented in Fig. 9.

On the grounds of the filtered and synchronously averaged waveforms the basic measures of the vibration signal are determined. The selected measures of the signal determined for the reference model, for the worn differential gear and for the worn triple joint are listed in Table 3.



Fig. 7. Casual wear of the differential gear



Fig. 8. Wear of the triple joint



Fig. 9. Filtered and synchronously averaged waveforms for the reference model, worn differential gear and worn triple joint

Table 3. S	Signal measures	for the good	system, for the	worn differential	l gear and for the	e worn triple joi	int
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Signal measure	No defects (reference model)	Worn triple joint	Worn differential gear	
Peak-to-peak value [m/s <sup>2</sup> ]	2.5760	13.4061	5.0762	
RMS value [m/s <sup>2</sup> ]	0.6843	3.3428	0.9652	
Average value [m/s <sup>2</sup> ]	0.5502	2.8630	0.7591	

The comparison of the residual signals of the base model and low frequency vibration signal of the gearbox, with the signals recorded for the defected differential gear and triple joint, is presented in Fig. 10.

The selected measures of the residual signals determined for the not defected drive system, for the worn out triple joint and for the worn out differential gear are listed in Table 4.

962. ON-BOARD DIAGNOSTICS OF MECHANICAL DEFECTS OF THE VEHICLE DRIVE SYSTEM BASED ON THE VIBRATION SIGNAL REFERENCE MODEL. IWONA KOMORSKA, ANDRZEJ PUCHALSKI

1.0		6	5	1
	Residual measure	Not defected system	Worn triple joint	Worn differential gear
	Peak-to-peak value [m/s <sup>2</sup> ]	0.1867	4.0768	0.5918
	RMS value [m/s <sup>2</sup> ]	0.0359	1.3855	0.1661
	Average value [m/s <sup>2</sup> ]	0.0391	1.6134	0.1896
	Uncertainty of the model		0.0118	

Table 4. Residual measures of the good drive system and selected defects at 3000 rpm



**Fig. 10.** Residual signals compared for: a) not defected system and worn out triple joint, b) not defected system and worn out differential gear

## 6. Conclusions

1. It is shown that the following signals are needed for the detection of mechanical defects of the vehicle driving system: from one vibrations sensor attached to the gearbox housing, from the crankshaft position sensor (for the synchronization of vibration signals) and from the throttle position sensor (for the assessment of the mechanical load of the engine).

2. The vibration signal of the vehicle driving system for the constant rotational speed of the engine can be treated as a stationary and ergodic signal – provided, that there are no abrupt load changes.

3. The vibration signal – based diagnostic autoregressive model (AR) of the vehicle driving system was proposed in the paper. In dependence of the subsystem being under diagnostics, the vibration signal must be properly prepared, which means that it should be filtered, synchronically averaged, or its autocorrelation function should be determined. The filter band selection – matching the model of the diagnosed subsystem – is also important.

4. This is an adaptive model, which means that its identification and diagnostics procedure is performed on-board in a way imperceptible for the driver.

5. The method of the on-line determination of the model coefficients was presented and its verification on the example of wearing out of the gearbox teeth, differential gear and triple joint was performed. For the selected defects the measures of vibration signals were determined and the residual signals were generated. On the basis of residual signals generated for the model and for the driving system in a good technical condition the model uncertainty was calculated.

6. Autoregressive models determined for the autocorrelation function of the vibration signal and for vibrations synchronously averaged were considered. In the first case the model is simpler (less coefficients) and is characterized by a smaller uncertainty, but using the normalized autocorrelation function it is only possible to state that the worn element is either in a good or bad state. In the second case the model has more coefficients, but it allows us to differentiate the wear state of the tested subsystem.

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