

bp

Tool wear monitoring in milling with neural networks by using motor current measuring

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Abstract

Indirect measuring of tool wear without stopping the process is a purpose in machining. In this paper an intelligent system for on-line tool wear estimation in milling through measuring spindle motor current is proposed. For this the current of motor in different condition of machining (feed, depth of cut and rpm of tool) and wear were measured with practical experiments and the effects of milling tool wear on current of motor is analysed. Based on the results, a back propagation (bp) neural network is developed and trained. Using this network the tool wear could be estimated while machining in different conditions with measuring current of motor. This system could be used in controlling and monitoring of the machining process.

Key words: tool wear, milling, neural networks, monitoring, motor current.

(unmanned Flexible UFMS

Manufacturing System)

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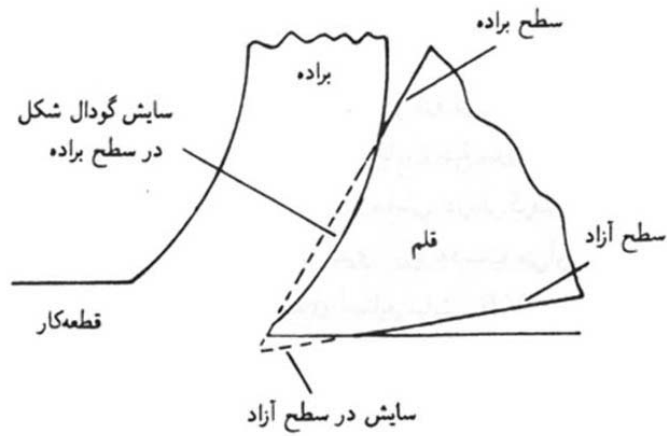
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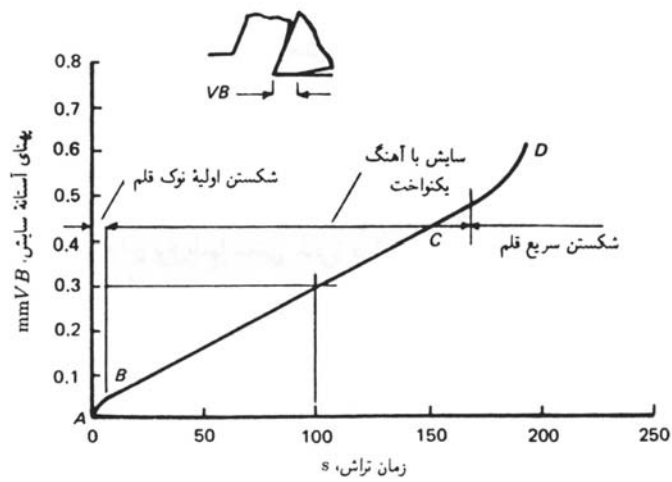
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$$V_B \quad h \quad V_C \text{ (NOTCH)}$$

$$W \quad w' \quad V_N$$

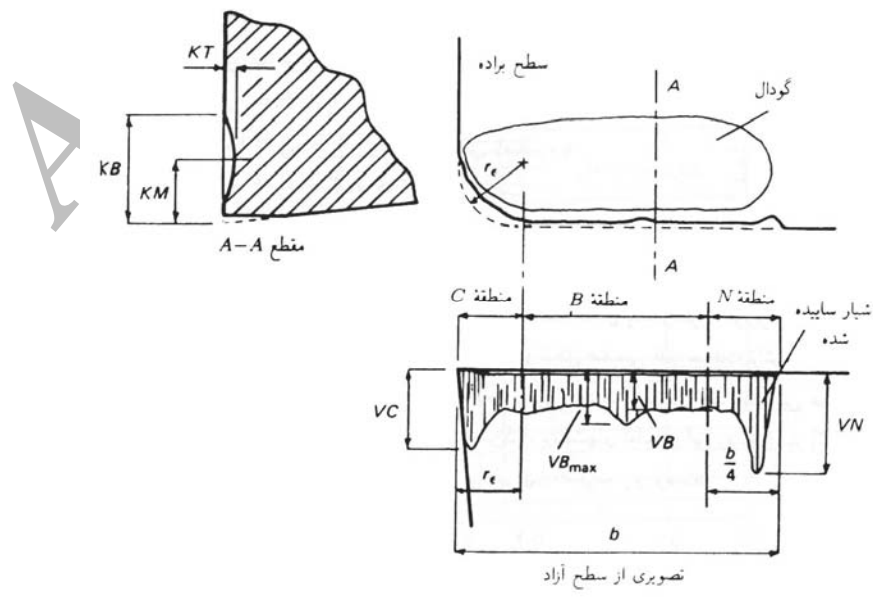
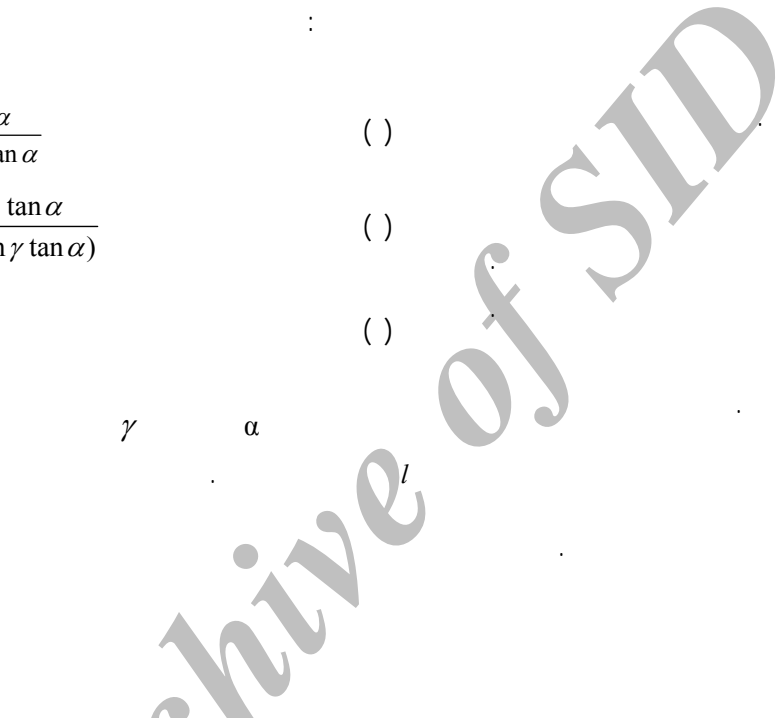
$$h = \frac{V_B \tan \alpha}{1 - \tan \gamma \tan \alpha} \quad ()$$

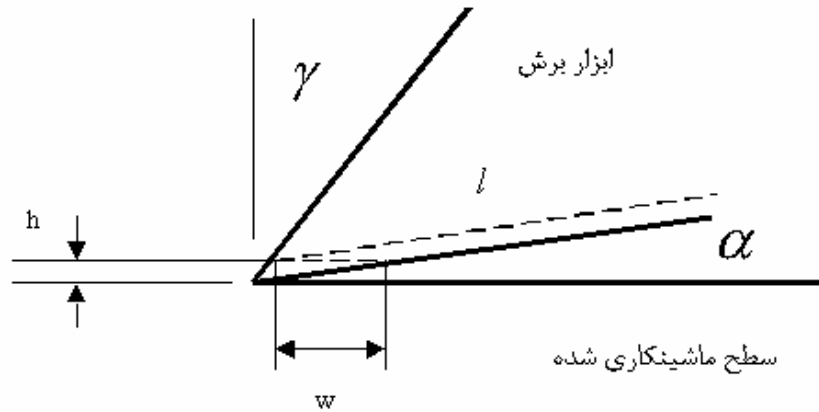
$$W = \frac{b V_B^2 \tan \alpha}{2(1 - \tan \gamma \tan \alpha)} \quad ()$$

$$W' \cong l b \sin \alpha \quad ()$$

$$V_B \quad V_{Bmax}$$

$b \quad \gamma \quad \alpha$





$$\begin{matrix} X_{th} & R_{th} \\ : & \\ & Z_{th} \end{matrix}$$

$$Z_{th} = R_{th} + iX_{th}$$

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$$\omega_{syn} = \frac{120F}{60p}$$

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p F

W mech

(s)

$$s = \frac{\omega_{syn} - \omega_{mech}}{\omega_{syn}}$$

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$$P_{mech} = T_{mech} \cdot \omega_{mech} = I_2^2 \frac{R_2}{s} (1-s)$$

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R_2 I_2

$$T_{mech} = \frac{1}{\omega_{syn}} * \frac{v_{th}^2}{(R_{th} + \frac{R'_2}{s})^2 + (X_{th} + X'_2)^2} * \frac{R'_2}{s}$$

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$$T_{mech} = \frac{1}{\omega_{syn}} I_2'^2 \frac{R'_2}{s}$$

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X_1 R_1 v_{th}
 X_2 R_2

I_2'

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$$f(x) = \frac{2}{1 + \exp(-x)} - 1$$

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$$f'(x) = \frac{1}{2} [1 + f_2(x)] [1 - f_2(x)]$$

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50mm ST37

Iso40-22-E

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mm

mm/min

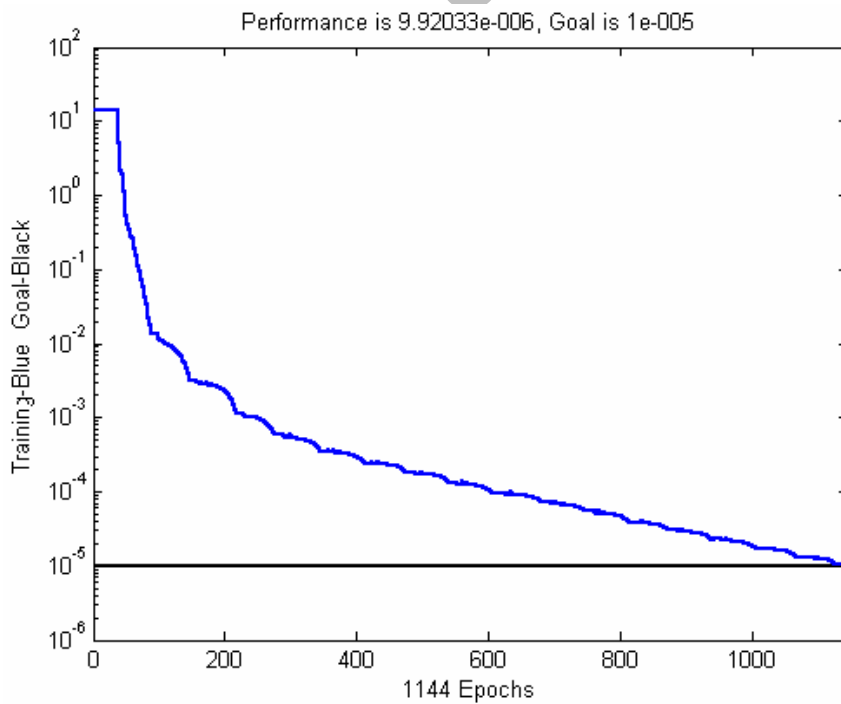
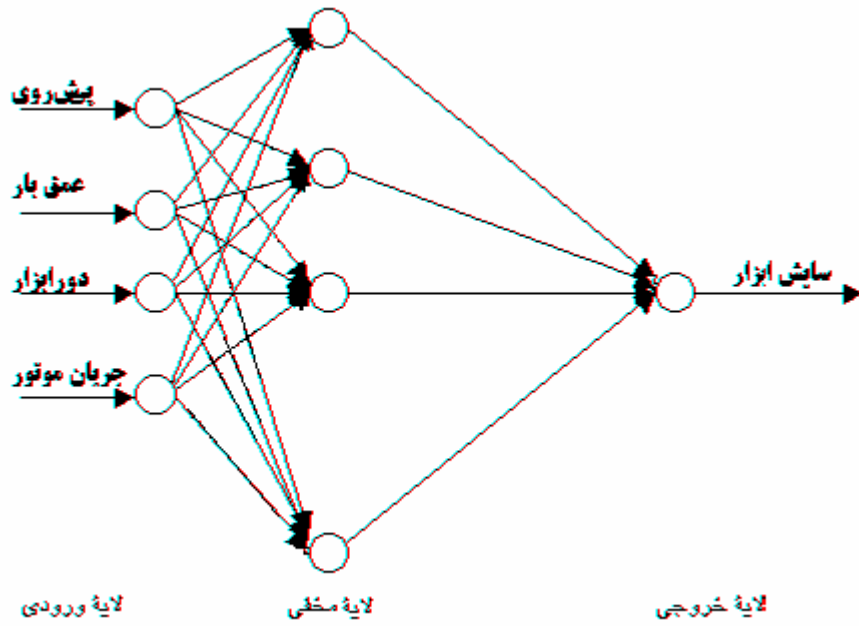
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(rpm

mm / /

()

-				(rpm) a=1.0 f=112
-	/	/	/	
-	/	/	/	
-				(max/min) a=1.0 n=500
-	/	/	/	
-	/	/	/	
/	/	/	/	(mm) n=500 f=112
/	/	/	/	
/	/	/	/	



cos (f_i)		a
:		b
$p = VI \cos(f_i)$	()	f
.	f_i	f_i
.		I
f_i		I_2
.		n
.		P
f_i		R_1
.		R_2
.		s
.		T
.		X_1
.		X_2
.		V
.		V_B
.		V_{th}
.		W
.		α
.		γ
.		ω
.		/
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