Adaptive Control of Active Magnetic Bearing against Milling Dynamics

Rong-Mao Lee * and Tsung-Chia Chen

Abstract: For improving the defects in milling processes caused by traditional spindle bearings, e.g., the dimensional discrepancy of a finished workpiece due to bearing wear or oil pollution by lubricant, a novel embedded cylindrical-array magnetic actuator (ECAMA) is designed for milling applications. Since ECAMA is a non-contact type actuator, a control strategy named fuzzy model-reference adaptive control (FMRAC) is synthesized to account for the nonlinearities of milling dynamics and magnetic force. In order to ensure the superior performance of spindle position regulation, the employed models in FMRAC are all constructed by experiments. Based on the experimental results, the magnetic force by ECAMA is much stronger than that by the traditional active magnetic bearing (AMB) design under the same test conditions and identical overall size. The efficacy of ECAMA to suppress the spindle position deviation with the aid of FMRAC has been verified as well via numerical simulations and practical metal cutting.

Keywords: active magnetic bearing; milling; adaptive control

1. Introduction

1.1. Active Magnetic Bearing (AMB)

Active magnetic bearing (AMB) is a magnetic actuator. In the past two decades, AMB has been gradually adopted as a non-contact, lubrication-free, and levitation-efficient device in various industrial applications. Owing to its merits, such as the capability to regulate spindle dynamics, AMB has become a potential element for high-speed rotating machinery, e.g., milling machines [1–3], gas turbines [4,5], turbo-molecular pumps [6], and so on.

There are two common configurations of AMB, i.e., radial design and modified radial design, as shown in Figure 1 [7]. Radial design is also called the planar type because all of the EM (electromagnetic) poles are placed on the same cross-plane. Although the axial length of the radial design is relatively short, the coil turns wound on individual EM poles are very limited. Since the magnetic force is induced by the coils, the force intensity of the radial design AMB is, therefore, restrained. To enhance the magnetic force, the modified radial design was proposed [8]. The EM poles of each magnetic flux loop in radial design (see the dotted line in Figure 1a) are re-arranged to make all the magnetic flux passing along the axial direction (see the dotted line in Figure 1b). Since the number of EM poles in each cross-plane of the modified radial design is merely a half of that of the planar design, the interior space between adjacent poles is significantly increased so that more coil turns can be wound on each yoke to increase the intensity of the magnetic force.
AMB is a promising technology for machining on account of low wear and friction, low maintenance cost, and long operating life. Park et al. [9] reported the design, analysis, and experimental results of a miniature spindle using magnetic bearings for very high-speed micromachining. The spindle is controlled with a DSP-based controller to evaluate the ability of tool orbit control under a spindle speed of 200,000 RPM. The control currents applied to AMBs allow them to not only stabilize the supported spindle but also to actively suppress chatter in the milling process. Gourc et al. [10] proposed a dynamical model of an active magnetic bearing spindle (AMBS) to identify the machining stability of high speed milling (HSM). The AMB and the servo, the flexible mode of rotor, the regenerative effect, and the nonlinearity when the tool leaves the cut under large vibrations are all taken into consideration in this model for the prediction of HSM stability. A sensitivity study of controller performance has been accomplished in this research as well. Another issue regarding the stability of the rotor-AMB system during acceleration/deceleration has been investigated by Huang et al. [11]. Since the rotational speed variation or acceleration/deceleration may destroy the stability of rotor-AMB system, an adaptive control strategy is designed to stabilize the system and to suppress chatter simultaneously. In addition to eliminating machining vibration, AMBs are also superior in the regulation of spindle position. A study of tool position control with the aid of AMBs during HSM was introduced by Smirnov et al. [12]. An improvement in the determination of surface micro-topography characteristics has been reported by Kim et al. [13]. The electromagnetic actuator (EMA) is employed to compensate for the spindle run-out error and the thermal displacement. An advanced EMA-spindle system has then been proposed for the purpose of generating a micro-pattern on a fixed work surface [14]. A proportional-integral-derivative (PID) controller and adaptive feed-forward cancellation are both adopted to stabilize the system and to compensate for the shaft run-out, respectively. Similar research of spindle control or chatter suppression with the aid of AMBs have also been published in the past five years [15–17].

1.2. Milling Dynamics

For milling applications, AMB is employed to counterbalance the cutting force during the milling process. Therefore, the control performance of AMB is highly determined by the modeling accuracy of milling dynamics. However, many parameters are involved in the milling operation; the dynamics of the cutting force are, therefore, complex and hardly predicted. For the purpose of retaining the MRR (Material Removal Rate) and prevent tool breakage for fixed-gain feedback controllers, an experimental approach was reported to build the dynamic milling model under various cut depths and spindle speeds [18]. Since the milling process is a typical time-varying system due to the variations of cutting conditions, a time-domain model was proposed for the analysis of cutting dynamics and chatter stability criterion. The cutting force is estimated by Oxley’s predictive machining theory according to the instantaneous un-deformed chip thickness, cutting parameters, material properties of the workpiece, and tool geometry [19]. Similarly, the first-order and second-order mathematical
models for milling dynamics were also proposed in 2002 and 2004, respectively [20,21]. A hybrid adaptive control algorithm is employed to tune the feedrate to regulate the variation of cutting force based on the first-order milling dynamic model [20]. Another second-order model was synthesized in a self-tuning regulator for milling processes under various cut depths [21]. Due to the complexity of milling operations, neural network [22], fuzzy logic algorithms [23], and heuristic methods [24] were all applied to the modeling of milling dynamics.

1.3. Embedded Cylindrical-Array Magnetic Actuator (ECAMA)

Though the two types of AMB designs shown in Figure 1 have ever been applied in computer numerical control (CNC) machines [1,3,6], the current AMB design still cannot meet the requirements for milling applications, such as the overall size of AMB and the force intensity [25–27]. For the purposes of solving these troublesome issues against milling applications, a novel embedded cylindrical-array magnetic actuator (ECAMA) is proposed [28]. The configuration of the ECAMA is shown in Figure 2. It is composed of modified concave-type yokes [8] and I-shaped electromagnets. The I-shaped electromagnet consists of an I-shaped silicon steel core and the coil. Since the coil is no longer wound on the yoke (please refer to Figure 1), the coil turns can be greatly increased. On the other hand, the overall size of the ECAMA in the radial direction can certainly be reduced. The top view of the ECAMA and spindle is shown in Figure 3. The yokes and the outer layer of spindle are all made of silicon steel to play the pathway of magnetic flux. The active magnetic milling module is depicted in Figure 4. The spindle position deviation is regulated by the magnetic force from the ECAMA. In addition, a self-sensing module is equipped close to the cutter to measure the spindle position deviation online.

![Configuration of the embedded cylindrical-array magnetic actuator (ECAMA).](image)

**Figure 2.** Configuration of the embedded cylindrical-array magnetic actuator (ECAMA).

![Top view of the ECAMA and spindle.](image)

**Figure 3.** Top view of the ECAMA and spindle.
2. Control Strategy for ECAMA against Milling Dynamics

2.1. Control Goal

Although a conventional PID controller can cooperate with AMBs to regulate the spindle position deviation on most occasions [29], for milling applications the PID controllers may lead to failure once the operating condition is changed rapidly. An alternative solution is to include a PID gain scheduling algorithm inside the controller, such as the cooperation between the PID controller and fuzzy logic algorithm [30], or logic-based supervisor strategy [31].

The block diagram of the milling process regulated by the ECAMA is shown in Figure 5, where the cutting force, \( F_c \), varies during milling operation and results in the spindle position deviation. On the other hand, the magnetic force, \( F_m \), is utilized as a compensatory force to regulate the spindle position. However, the milling dynamics are highly nonlinear and determined by several cutting operation parameters. As a result, the variation of cutting force is, hence, drastic and complex [3]. Similarly, the magnetic force by the ECAMA is highly nonlinear, as well. In view of the arguments above, the FMRAC (fuzzy model-reference adaptive control) is developed in this work to account for the nonlinearities of milling dynamics.

Adaptive control can operate without the need of prior information about the boundaries of uncertainties or time-varying parameters [32,33]. Since the dynamics of the cutting force and magnetic force are both highly nonlinear, a control strategy composed of reference models and fuzzy logic algorithms is designed to regulate the spindle position deviation. For reducing the control error, the employed mathematical models are going to be constructed by experiments instead of a theoretical approach. After the dynamic models are built, the variation in cutting force can be numerically estimated via a fuzzy logic algorithm according to the current operation conditions and parameter settings.
2.2. Fuzzy Model-Reference Adaptive Control (FMARC)

The block diagram of the fuzzy model-reference adaptive control (FMARC) strategy is shown in Figure 6. The fuzzy milling dynamic models are constructed by experiments for the purpose of estimating the cutting force, \( F_c \) [3] online. Once the estimated cutting force, \( \hat{F}_c \), is obtained according to the present axial cut depth, \( a_C \), feedrate, \( f \), and spindle speed, \( \Omega \), the compensatory magnetic force, \( F_m \), can be determined to counter-balance \( F_c \), i.e., \( F_m = -\hat{F}_c \). Similarly, the reference model of spindle dynamics is also built by experiments. On the other hand, the control inputs provided by power amplifiers to the ECAMA are nothing but electric currents. In other words, the control commands to the ECAMA are coil currents, instead of magnetic force requirements. Since the nonlinear magnetic force is a function of coil current and air gap, the characteristics of the ECAMA can also be explicitly revealed by experiments [28]. By employing the fuzzy milling dynamic model and the magnetic force model, the corresponding coil current for required compensatory magnetic force can be determined by the controller.

Although all the employed dynamic models are constructed by experiments, these models are just approximate copies and impossible to completely represent the actual dynamics of milling operation and magnetic force. As a result, a certain degree of estimation error might exist between the actual compensatory force, \( F_m \), and the actual cutting force, \( F_c \), due to the inaccuracy of modeling. In order to enhance the performance of spindle position regulation, a fuzzy compensator is proposed to modulate the control command via adjusting the feedback gain of controller. The rules for fuzzy compensator are based on the error, \( e_2 \), between the compensatory force determined by controller, \( \hat{F}_m \), and the compensatory force evaluated by the reference model of spindle dynamics, \( F_m \). Assume there are \( \ell \) terms of rules in the rule base and the rule \( L^i \) can be expressed as follows:

\[
L^i : \quad \text{if } x(k) \text{ is } D^i_1, x(k-1) \text{ is } D^i_2, \ldots, x(k-n+1) \text{ is } D^i_n \text{ then } x(k+1) = a^i_1 x(k) + a^i_2 x(k-1) + \cdots + a^i_n x(k-n+1) \quad (1)
\]

where \( i = 1, 2, \ldots, \ell, D^i_j \) is the fuzzy set and \( a^i_j \) is constant. Rule \( L^i \) can be further rewritten as:

\[
X(k+1) = M_i X(k) \quad (2)
\]

where:

\[
X(k+1) = \begin{bmatrix} x(k+1) & x(k) & \cdots & x(k-n+2) \end{bmatrix}^T \quad (3a)
\]

\[
X(k) = \begin{bmatrix} x(k) & x(k-1) & \cdots & x(k-n+1) \end{bmatrix}^T \quad (3b)
\]
The defuzzification of Equation (2) can be stated as follows:

\[ X(k+1) = \frac{\sum_{i=1}^{l} w_i M_i X(k)}{\sum_{i=1}^{l} w_i} \]  \hspace{1cm} (4)

where \( w_i \) is the fitness function for rule \( L_i \) and is defined as follows:

\[ w_i = \min_j (D_i(x)), \quad \omega_i \geq 0 \]  \hspace{1cm} (5)

Figure 6. Block diagram of the fuzzy model reference adaptive control system.

The membership functions for fuzzification are shown in Figure 7, where the horizontal and vertical axes are corresponding to the normalized error, \( E_2 \) and its derivative, \( \dot{E}_2 \) respectively.

Remarks:

- \( d_i \): Reference Spindle Position Deviation (SPD)
- \( \Omega \): Spindle Speed
- \( u \): Controller Output
- \( f \): Feedrate
- \( a_c \): Axial Cut Depth
- \( P_m \): Induced Compensatory Force by ECAMA
- \( d \): Measured SPD

The membership functions for fuzzification are shown in Figure 7, where the horizontal and vertical axes are corresponding to the normalized error, \( E_2 \) and its derivative, \( \dot{E}_2 \) respectively. The
obtained fuzzy variables, $e_2$ and $\dot{e}_2$ are firstly normalized within the interval, $[-1,1]$, and it ends up with seven triangular membership functions for all individual variables.

The complete rule base for the membership functions is listed in Table 1. The fuzzification plane in Figure 7 can be further divided into several sectors. The symbols, $u_{ij}$, where $i = 1 \sim 7$, $j = 1 \sim 7$, at the four corners of each rectangular grid are the default output rules for defuzzification. The membership functions for the decision of output rule is shown in Figure 8, where the horizontal and vertical axes are the output and the weight coefficient respectively. In order to explain how the weight coefficients are evaluated, the rectangular grid of slash in Figure 7 can be further represented by Figure 9. The four corners of this rectangular grid are replaced by points $S \sim V$. Then the weight coefficients corresponding to the associated rectangular grid can be expressed as follows:

\begin{align}
    w_{S1} &= \frac{(m + L_f - s)}{L_f} \\
    w_T &= 1 - w_{S1} \\
    w_{S2} &= \frac{(n + L_f - t)}{L_f} \\
    w_V &= 1 - w_{S2}
\end{align}

Finally, the output of the fuzzy compensator, $u_F$, can be expressed by applying sum-product method:

\[ u_F = \frac{\mu_1 u_{11} + \mu_2 u_{12} + \mu_3 u_{21} + \mu_4 u_{22}}{\mu_1 + \mu_2 + \mu_3 + \mu_4} \]

where:

\[ \mu_1 = w_{S1} w_V, \quad \mu_2 = w_T w_V, \quad \mu_3 = w_{S1} w_{S2}, \quad \mu_4 = w_T w_{S2}. \]

Following the steps stated above, the adaptive gain can, thus, be tuned by the fuzzy compensator.
Figure 8. Membership functions for decision of output rules.

Figure 9. Rectangular grid of the fuzzification plane.

Table 1. Rule base for fuzzy compensator.

<table>
<thead>
<tr>
<th>$\epsilon_2/\epsilon_2$</th>
<th>NB</th>
<th>NM</th>
<th>NS</th>
<th>Z</th>
<th>PS</th>
<th>PM</th>
<th>PB</th>
</tr>
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<td>NM</td>
<td>NB</td>
<td>NB</td>
<td>NB</td>
<td>NM</td>
<td>NS</td>
<td>Z</td>
<td>PS</td>
</tr>
<tr>
<td>NS</td>
<td>NB</td>
<td>NB</td>
<td>NM</td>
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<td>PS</td>
<td>PM</td>
</tr>
<tr>
<td>Z</td>
<td>NB</td>
<td>NM</td>
<td>NS</td>
<td>Z</td>
<td>PS</td>
<td>PM</td>
<td>PB</td>
</tr>
<tr>
<td>PS</td>
<td>NM</td>
<td>NS</td>
<td>Z</td>
<td>PS</td>
<td>PM</td>
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<tr>
<td>PM</td>
<td>NS</td>
<td>Z</td>
<td>PS</td>
<td>PM</td>
<td>PB</td>
<td>PB</td>
<td>PB</td>
</tr>
<tr>
<td>PB</td>
<td>Z</td>
<td>PS</td>
<td>PM</td>
<td>PB</td>
<td>PB</td>
<td>PB</td>
<td>PB</td>
</tr>
</tbody>
</table>

Remarks: NB—negative big, NM—negative medium, NS—negative small, Z—zero, PS—positive small, PM—positive medium, PB—positive big.

2.3. Performance of the FMRAC

In order to demonstrate the performance of the proposed FMRAC strategy, the PID controller is compared as the benchmark. The workpiece profile for numerical simulation is depicted in Figure 10 where the dotted line is the cutting path to be undertaken. That is, the cutter will encounter various resistances in the forms of ramp (Section II) and step (Section IV) during milling. The parameters of the milling operation and ECAMA/spindle system for numerical simulations are listed in Table 2.

The simulation results are shown in Figures 11–14. The required coil currents in $X$- and $Y$-axes under various control laws are shown in Figures 11 and 12 respectively. The variations of coil currents with respect to PID and FMRAC strategies in $Y$-axis are similar. However, the current variation under
PID control in the feed direction (X-axis) is much more drastic in comparison with that under the FMRAC strategy. On the other hand, the maximum current under the FMRAC strategy is also lower than that under PID control especially during step resistance (referred to Section IV in Figure 10). The related spindle position deviation in X- and Y-axes are shown in Figures 13 and 14 respectively. Significant tremble of position deviation in X-axis can be observed under PID control, no matter ramp (Section II in Figure 10) or step (Section IV in Figure 10) cutting is engaged. In addition, the position deviations in Y-axis under both of the two control strategies are relatively smoother than those in X-axis due to the design of slot milling simulation. The proposed FMRAC strategy is verified by numerical simulation for its efficacy of spindle position regulation. In comparison with PID control, the performance of spindle position regulation under the proposed FMRAC strategy is much improved and the required coil current is also reduced.

Table 2. Parameters of milling operation and ECAMA/spindle system for numerical simulations.

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Nominal air gap between spindle and yoke</td>
<td>$10^{-3}$ m</td>
</tr>
<tr>
<td>Distance between flexible coupling and top yoke</td>
<td>$5 \times 10^{-2}$ m</td>
</tr>
<tr>
<td>Distance between flexible coupling and bottom yoke</td>
<td>0.2 m</td>
</tr>
<tr>
<td>Distance between flexible coupling and gap</td>
<td>0.23 m</td>
</tr>
<tr>
<td>Distance between flexible coupling and cutter tip</td>
<td>0.25 m</td>
</tr>
<tr>
<td>Transverse mass moment of inertia of spindle</td>
<td>$2.2572 \times 10^{-2}$ Kg - m^2</td>
</tr>
<tr>
<td>Polar mass moment of inertia of spindle</td>
<td>$5.9 \times 10^{-4}$ Kg - m^2</td>
</tr>
<tr>
<td>Stiffness of current</td>
<td>9.54 N/A</td>
</tr>
<tr>
<td>Stiffness of displacement</td>
<td>$-14313$ N/m</td>
</tr>
<tr>
<td>Spinning speed of spindle</td>
<td>10k rpm</td>
</tr>
<tr>
<td>Feedrate</td>
<td>3 m/min</td>
</tr>
</tbody>
</table>

Figure 10. Profile of workpiece for numerical simulation.

Figure 11. Comparison of required coil current for the ECAMA in the X-axis under various control laws.
3. Experimental Setup and Results

The prototype specifications of milling module and ECAMA are as follows:

- Axial length of milling module: 40 cm.
- Maximum radial length of milling module: 11 cm.
- Total weight of milling module: 7.5 kg.
- Power of drive motor: 1 kW.

The comparison between the ECAMA and traditional AMB is listed in Table 3. It is obvious that the ECAMA is able to provide a much stronger regulation force than that of a traditional design of similar size.

<table>
<thead>
<tr>
<th>Specifications</th>
<th>ECAMA</th>
<th>Modified Radial Design</th>
<th>AMB</th>
</tr>
</thead>
<tbody>
<tr>
<td>Diameter (mm)</td>
<td>88</td>
<td>85</td>
<td></td>
</tr>
<tr>
<td>Axial Length (mm)</td>
<td>122</td>
<td>81</td>
<td></td>
</tr>
<tr>
<td>Coil Turns</td>
<td>1200</td>
<td>220</td>
<td></td>
</tr>
<tr>
<td>Max Induced Magnetic Force (N)</td>
<td>224</td>
<td>65 (3 A coil current supplied)</td>
<td></td>
</tr>
</tbody>
</table>

Remark: The diameter of wound coil is 0.5 mm for both ECAMA and modified radial design AMB.

The test rig, including the milling machine (How-Mau CNC Machinery Co., Ltd., New Taipei City, Taiwan), is depicted in Figure 15. Commercial software MATLAB/Simulink (R2011a, The MathWorks, Natick, MA, USA, 2011), and signal processing interface (DS1104, dSPACE, Paderborn, Germany, 2007), have been also employed in the intensive experiments. A HSS (high-speed steel) cutter (Chien Wei Machinery&Hardware Co., Ltd., Taichung City, Taiwan) of 8 mm diameter and four cutting blades is equipped and the cutting path is defined in Figure 16. The operation parameters for milling test are set as follows:

- Materials of workpiece: acrylic, aluminum and copper (three types of materials).
- Spindle speed ($\omega$): 3000 rpm
Maximum speed of drive motor: 24k rpm
Total weight of ECAMA: 2.8 kg
Axial length of ECAMA: 12.2 cm
Maximum radial length of ECAMA: 10 cm
Maximum power consumption of ECAMA: 60 W (by an individual I-shape electromagnet)

The comparison between the ECAMA and traditional AMB is listed in Table 3. It is obvious that the ECAMA is able to provide a much stronger regulation force than that of a traditional design of similar size.

Table 3. Comparison between ECAMA and traditional AMB design.

<table>
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<tr>
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<td>88</td>
<td>85</td>
</tr>
<tr>
<td>Axial Length (mm)</td>
<td>122</td>
<td>81</td>
</tr>
<tr>
<td>Coil Turns</td>
<td>1200 (on each individual I-shape silicon steel core)</td>
<td>220 (on each individual pair of electromagnetic poles)</td>
</tr>
<tr>
<td>Max Induced Magnetic Force (N)</td>
<td>224 (1.5 A coil current supplied)</td>
<td>65 (3 A coil current supplied)</td>
</tr>
</tbody>
</table>

Remark: The diameter of wound coil is 0.5 mm for both ECAMA and modified radial design AMB.

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- Materials of workpiece: acrylic, aluminum and copper (three types of materials).
- Spindle speed (Ω): 3000 rpm
- Feedrate (f): 300 mm/min.
- Axial cut depth (Da): 2.5 mm.
- Radial cut depth (Dr): 1 mm.

Figure 15. Test rig for milling operation.
The realistic trajectories of spindle position deviation in X-Y plane for three kinds of materials are shown in Figures 17–19 respectively. The maximum spindle position deviations for the three different kinds of workpieces are similar, but the spindle dynamics under acrylic cutting is much more complex than those under other materials. Since acrylic is brittle, the variation of cutting force is possibly more drastic than that of metal cutting. Therefore, the compensative force by the ECAMA during acrylic cutting should be tuned rapidly according to the cutting condition. In addition, though the trajectories of spindle position deviation corresponding to aluminum and copper are similar, the spindle position deviation for copper cutting is smoother than that for aluminum cutting, due to different metal properties (i.e., copper is more ductile). The photographs of the finished workpiece surface are shown in Figures 20–22. The resulting ridges on aluminum are more significant than those on acrylic and copper. The efficacy of the proposed FMRAC for spindle position regulation is verified by realistic milling.
tools; R.-M. Lee wrote the paper.

R.-M. Lee conceived and designed the experiments; R.-M. Lee performed the experiments; R.-M. Lee and T.-C. Chen analyzed the data; T.-C. Chen contributed reagents/materials/analysis tools; R.-M. Lee wrote the paper.

Author Contributions:

4. Conclusions

A fuzzy compensator is designed to achieve the adaptive gain control. As previously mentioned, the estimation errors are hard to be avoided. A fine-tuning scheme for control gain is, therefore, embedded to diminish the degree of estimation error. For further applications, such as chatter suppression, the membership functions and the decision rules have to be addressed carefully.

Figure 18. Trajectory of spindle position deviation in X-Y plane (workpiece: aluminum).

Figure 19. Trajectory of spindle position deviation in X-Y plane (workpiece: copper).

Figure 20. Micrograph of finished workpiece (acrylic).

Figure 21. Micrograph of finished workpiece (aluminum).
4. Conclusions

In this work, the control strategy for the ECAMA to regulate spindle position deviation against milling dynamics has been proposed and verified. According to the simulation and experimental investigations, the following points are made:

1. The ECAMA is a potential solution for milling applications of AMB. It is designed to deal with the issues of magnetic force intensity and overall AMB size at the same time. That is, the regulation force by AMB can be much enhanced under a reasonable size. However, for regular industrial machinery, the size of the ECAMA should be much larger than that in this work to achieve the requirements of industrial applications.

2. Since the dynamics of milling processes and magnetic force are both highly nonlinear, estimation errors of the cutting force or magnetic force are definitely met. These force dynamics are investigated by experiments in this work instead of parameters analysis by theoretical methods. The accuracy of experimental models is highly determined by the experiment design. Therefore, the test rig design and the selection of analysis method should be addressed carefully.

3. A fuzzy compensator is designed to achieve the adaptive gain control. As previously mentioned, the estimation errors are hard to be avoided. A fine-tuning scheme for control gain is, therefore, embedded to diminish the degree of estimation error. For further applications, such as chatter suppression, the membership functions and the decision rules have to be modified to make the regulation behavior faster and more efficient.

Author Contributions: R.-M. Lee conceived and designed the experiments; R.-M. Lee performed the experiments; R.-M. Lee and T.-C. Chen analyzed the data; T.-C. Chen contributed reagents/materials/analysis tools; R.-M. Lee wrote the paper.

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