Repetitive control of an XYZ piezo-stage for faster nanoscanning: numerical simulations and experiments

by

Selman A. Cebeci

A Thesis Submitted to the

Graduate School of Engineering in Partial Fulfillment of the Requirements for

the Degree of

Master of Science

in

Mechanical Engineering

Koc University

March 2011

Koc University

Graduate School of Sciences and Engineering

This is to certify that I have examined this copy of a master's thesis by

Selman A. Cebeci

and have found that it is complete and satisfactory in all respects, and that any and all revisions required by the final examining committee have been made.

Committee Members:

Çağatay Başdoğan, Ph. D. (Advisor)

Yaman Arkun, Ph. D.

Levent Güvenç, Ph. D.

Date:

ABSTRACT

A Repetitive Controller (RC) is implemented to control the z-axis movements of a piezo-scanner used for AFM scanning and then tested through scan experiments and numerical simulations. The experimental and simulation results show that the RC compensates phase delays better than the standard PI controller at high scan speeds, which leads to less scan error and lower interaction forces between the scanning probe and the surface being scanned. Since the AFM experiments are not perfectly repeatable in the physical world, the optimum phase compensators of the RC resulting this performance are determined through the numerical simulations performed in MATLAB/Simulink. Furthermore, the numerical simulations are also performed to show that the proposed RC is robust and does not require retuning of these compensators when the consecutive scan lines are not similar and a change occurs in the probe characteristics.

ÖZET

Bu tezde, Atomik Kuvvet Mikroskobunda kullanılan piezo tarayıcının z-ekseni için RC uygulanmış, simülasyonlar ve tarama deneyleriyle test edilmiştir. Simülasyonlar ve tarama deneyleri tekrarlı kontrolcünün standart olarak kullanılan PI kontrölcüye göre faz gecikmelerini daha iyi telafi edebildiğini ve sonuç olarak yüksek hızlarda daha az hata ve daha düşük kuvvet ile tarama yapabildiğini göstermiştir. AKM deneylerinde tekrarlanabilirlik düşük olduğu için simülasyonlar tekrarlı kontrolcü için gerekli olan faz telafisinin bulunması için kullanılmıştır. Simülasyonlar ayrıca kontrollü olarak oluşturulması mümkün olmayan fakat deneyler sırasında rastlanabilecek durumlar için tekrarlı kontrolcünün gürbüzlüğünü test etmek ve performansını PI kontrolcü ile karşılaştırmak amacıyla kullanılmıştır.

ACKNOWLEDGEMENTS

I would state my sincere gratitude to my advisor, Çağatay Başdoğan for his guidance throughout my study. His critical criticism along with his positive attitude was the main contributor to my thesis. I would also like to thank Prof. Levent Güvenç and Prof. Yaman Arkun for their participation in the thesis committee.

I am thankful to Serkan Necipoğlu for his help on my research and friendship during the long lab hours. I would express my warmest thanks to Yunus Emre Has, for being my colleague and more importantly my buddy.

I wish to thank Can Gökgöl and Emre Ölçeroğlu for being my office mates. Their friendship made my graduate education more fun.

I appreciate IEEE Explore for providing me with the academic knowledge I have and having a simple search interface.

Finally, I wish to express my deepest thanks to my wife, Betül for her love and encouragement. Her continuous support made the completion of my thesis much easier.

TABLE OF CONTENTS

LIST OF FIGURES

Chapter 1

INTRODUCTION

Increasing the scan speed of an AFM without sacrificing the scan quality is an active area of research. One method of increasing the scan speed is to improve the mechanical design of the AFM while the other method is to replace the conventional PI controller used in the feedback loop with a more sophisticated one. In this study, we focus on the latter method and propose a repetitive controller (RC) for faster scanning of nano-scale surfaces in tapping mode AFM. The existing control studies on AFM can be categorized into two groups: a) those aiming to control the scanning probe and b) those focusing on the control of the scanner (i.e. the piezo actuator holding the sample). The latter group can be further divided into two sub-groups: controllers designed to improve the scan speed by adjusting the periodic lateral movements of the scanner on X-Y plane and those designed to maintain the image quality at high scan speeds by adjusting the vertical movements of the scanner along the Z-axis. In Figure 1.1, the movements of the stage along the X, Y and Z axes during an AFM scan are illustrated. Typically, the stage is moved back and forth on the Xaxis by sending a triangular voltage signal to the X-actuator while it is slowly advanced along the Y axis by sending a ramp voltage signal to the Y-actuator. The resulting scan motion is a triangular trajectory on the X-Y plane. Obviously, tracking a periodic back and forth motion is more challenging than tracking a ramp input. Hence, the research in that area is mostly focused on improving the scan speed along the X-axis. Moreover, the controllers used for the lateral axes (X and Y) aim to reduce the image distortions that occur due to the nonlinear behavior of the piezo-actuators such as creep, hysteresis, and thermal effects as well as the coupling between them.

Figure 1.1. a) Illustration of the AFM setup b) The movements of the stage during the scanning of nano-scale steps.

Typically, a PI controller is used for controlling the X-axis motion in commercial AFMs. But due to the bandwidth limitation of a standard PI controller, making an improvement in

scan speed is limited. For this reason, more sophisticated control techniques have been proposed to achieve higher scan speeds. In [1], inverted models of creep, hysteresis, and induced vibrations are utilized in the controller design to reduce the undesired nonlinear effects. A feedback controller combined with a feedforward controller is proposed in [2] and [3], where the feedback controller is used to compensate for the creep and hysteresis while the feedforward controller is used to increase the scan speed. In [4], Hung et al. integrate the feedback and feedforward controllers using a post-fitting control scheme. An inversion based iterative controller is also proposed in [5] to reduce the undesired hysteresis effects and in [6] to reduce the output oscillations via pre-filter. This approach is improved in [7], where a model-free design is suggested for easier implementation of the controller. An iterative learning controller (ILC) is used to compensate for hysteresiscaused positioning errors in AFM scanning [8]. In [9], an ILC is combined with a H_{∞} feedback controller and a ratio that determines the frequency range in which feedback or feedforward control should be employed is introduced. In [10], Bhikkaji et al. propose an integral resonant controller for suppressing the resonant mode of the piezo actuator to increase the scanning speed. An RC is proposed in [11] and [12] to increase the lateral scan speed by taking advantage of the repetitive nature of the trajectory followed by the nanopositioner on the X-Y plane.

Developing a controller for the Z-axis is more challenging than developing one for the X and Y axes since the trajectory followed on the X-Y plane is periodic and known in advance, whereas the sample surface to be scanned is an unknown disturbance affecting the Z-motion of the nanopositioner. Moreover, the interaction forces between the scanning probe and the surface being scanned are nonlinear, which makes it more difficult to adjust the probe position with respect to the sample surface along the Z-axis. Again, a standard PI controller is typically used for the Z-axis control of nanopositioners though more advanced control techniques have been suggested recently to improve the scan performance. In [13], Schitter et al. utilize an H_{∞} controller for the Z-axis piezo-actuator to keep the interaction forces between the probe and the sample at a constant value, while the movement of the scanner is simultaneously tracked by a model-based feedforward controller to reduce the scan error at high scan speeds. In [14], an adaptive PID controller is implemented for the Zaxis piezo-actuator to modify the PID gains on the fly during the scan process based on the control error. In [15], Fujimoto and Oshima propose a Surface Topography Learning Observer (STLO) to estimate the profile of a scan line in the forward pass and then use it during the backward pass of the same line to reduce the scan error. An ILC working in parallel with an H_{∞} feedback controller is developed in [16] to improve the dynamic response of the Z-axis piezo and provide robustness. In their implementation, the control input for tracking the current scan line is generated based on the tracking results of the previous scan line. The control theory behind an ILC is similar to that of an RC, but the ILC requires resetting of the initial conditions at the start of each iteration step for the implementation.

In this study, an RC for controlling the Z-axis movements of a piezo-stage is proposed. The RC proposed in [11] and [12] is similar to this one, but aims to control the lateral axes of a nanopositioner rather than its Z-axis. Controlling the Z-axis using an RC is more challenging than that of the lateral axes since the surface topography is constructed based on the vertical movements, which is an unknown exogenous disturbance for the system. An RC is known to work well for systems tracking repetitive inputs or rejecting repetitive disturbances. In AFM scanning, the profiles of the successive scan lines are repetitive since the advancement along the Y-axis is small during the raster scanning. The proposed RC takes advantage of the knowledge of the previous scan line while scanning the current one to achieve a better scan performance than that of a standard PI controller. In comparison to the earlier work in the same area [17], we further investigate the effect of our control design on the scan performance using numerical simulations performed in MATLAB/Simulink for various scan scenarios. Numerical simulations performed in MATLAB/Simulink have been used to investigate the effect of changing Q factor of the probe on the scan performance [18] and imaging parameters on the tip sample forces [19]. Following the design of our RC, the effect of tunable parameters (i.e. phase compensators) on the scan performance are investigated extensively using numerical simulations. Since

there are many factors (i. e. temperature, dust, humidity, ground vibration, electrical noise, etc) that affect the AFM scan in the real world settings, the scan results are not perfectly repeatable and an extensive analysis based on the scanning experiments is almost impossible. Numerical simulation is an alternative solution to this problem. In this study, the optimum phase compensators (i.e. tunable parameters) of the RC are determined through the numerical simulations performed in MATLAB/Simulink. Furthermore, the numerical simulations are performed to show that the RC does not require re-tuning of these compensators when a change occurs in the probe characteristics. After tuning the RC based on the results of the numerical simulations, its performance to that of the PI controller is compared through the scanning experiments performed in our AFM set-up.

Chapter 2

REPETITIVE CONTROLLER BASICS

RC is a powerful controller for tracking or rejecting repetitive signals. It is based on the internal model principle which states that the steady state error converges to zero if the generator of the applied reference input is placed on the forward path. A good example is an integrator used in front of a plant in a feedback control loop when the input signal is a step function. RC is simply a memory loop which acts as the generator of any repetitive input (Figure 2.1a). Therefore, it can track or reject any repetitive signal as long as its period, τ_d , is determined correctly. The effect of a memory loop can be better appreciated when the Bode plot in Figure 2.1b is inspected. As shown in the figure, the gain tends to infinity at the fundamental frequency and at its harmonics. Hence, if the input signal to be tracked or the disturbance signal to be rejected is repetitive, one can introduce high control gains to the system using an RC, thereby reducing steady state error to zero.

Figure 2.1. a) The memory loop b) The frequency response of the memory loop.

The block diagram of a typical RC [20] is shown in Figure 2.2. The design of an RC involves the design of two low-pass filters, Q(s) and B(s) along with the selection of two tunable parameters, time advances τ_q and τ_b. Q(s) is a low-pass filter with a unit DC gain, which is utilized to filter out the undesired dynamics at high frequencies and to maintain stability in the presence of uncertainties. $B(s)$ is a low-pass filter used for improving the performance and relative stability of the RC. Hence, the main purpose of using the Q-filter is to reduce the gain introduced by the RC to the system at the undesired harmonics, which would otherwise lead to instability due to the effects of noise and modeling uncertainties at high frequencies. The design of the Q-filter also affects the operating bandwidth of the RC. The additional phase introduced to the system by the Q-filter can be compensated by a small time advance τ_q (Figure 2.2). The main objective of using the B-filter is to improve the relative stability of the control system. Consider the regeneration spectrum defined for an RC by Srinivasan and Shaw [21]:

$$
R(\omega) = |Q(j\omega)\left[1 - B(j\omega)\frac{G(j\omega)}{1 + G(j\omega)}\right]|
$$
\n(2.1)

For a stable implementation, the regeneration spectrum, $R(\omega)$, should be less than 1 for all frequencies of ω [21]. Therefore, the B-filter is simply designed to invert the transfer function $G(j\omega)/[1+G(j\omega)]$ at the operating frequencies. The additional phase introduced by the B-filter and the plant can be also compensated by a small time advance, τ_b (Figure 2.2). In our implementation, the time advances, τ_q and τ_b , are tuned for the best performance using the numerical simulations. The effect of these parameters on the system performance is further discussed in Chapter 5.

Figure 2.2. A typical RC scheme.

Chapter 3

REPETITIVE CONTROL FOR AFM

3. 1. Experimental Setup

The components of the home-made AFM developed in [22] are as shown in Figure 3.1. In this set-up (Figure 3.1), an AFM probe mounted on a piezo-buzzer is brought close to the sample surface via a manual stage for coarse adjustment. The vibrational velocity of the probe tip actuated by the buzzer is measured through a laser doppler vibrometer (LDV). The RMS value of the velocity signal obtained from the vibrometer is calculated by an analog electronic circuit first, and then transferred to a computer via a data acquisition card. Some modifications were made on this set-up to implement an RC in series with a standard PI controller as shown in Figure 3.2. As a result, the tracking errors are corrected by the RC while the PI controller handles the errors caused by hysteresis, creep, and thermal drift occurring at low-frequencies. The software for the implementation of the controller is developed in LabVIEW.

Figure 3.1. The home-made AFM set-up.

The vibration amplitude of the probe (A_{act}) is calculated by integrating the RMS velocity, which is then subtracted from the set amplitude (A_{set}) to calculate the error (Figure 3.2). The error signal is then fed into the controller to generate the control signal, which is sent to the piezo-actuated XYZ nano-stage to adjust the tip-sample separation along the Z-axis. The movements of the nano-stage along the Z-axis are recorded at each grid point on the X-Y plane to obtain the topographical image of the sample surface.

Figure 3.2. Our implementation of the RC for AFM scanning.

3. 2. System Identification

An accurate model of the nano-stage is necessary for the proper implementation of the proposed RC. For this purpose, the time-dependent displacement response of the nanostage along the Z-axis to a step and an impulse input voltage is recorded. A transfer function model, $G'(z)$, is fit to the experimental data using the least squares method. The order of the polynomials in the numerator and the denominator are chosen to be 3 and 7 respectively. This choice is made by trial and error placing emphasis on the best fit with the minimum degree. The $G'(z)$ is of the following form:

$$
G'(z) = \frac{b_3 z^3 + b_2 z^2 + b_1 z + b_0}{z^7 - a_6 z^6 + a_5 z^5 - a_4 z^4 + a_3 z^3 - a_2 z^2 + a_1 z - a_0}
$$
(3.1)

The coefficients of the polynomials are calculated using the formula below:

$$
\alpha^{k+7} = [-\alpha^{k+6} \dots - \alpha^k u^{k+3} \dots u^k] \begin{bmatrix} a_6 \\ \dots \\ a_0 \\ b_3 \\ \dots \\ b_0 \end{bmatrix} + e^k
$$
(3.2)

\n
$$
\alpha^{k+7} = [-\alpha^{k+6} \dots - \alpha^k u^{k+3} \dots u^k] \begin{bmatrix} a_6 \\ \dots \\ a_9 \\ \dots \\ b_1 \end{bmatrix}
$$
(3.2)

Where $u(z)$ and $\alpha(z)$ represent the input and output signals, respectively. The transfer function coefficients are stored in the vector θ . The coefficients minimizing the curve fitting error, *e*, can be calculated with the following equation,

$$
\theta = (\varphi^* \varphi)^{-1} \varphi^* \lambda \tag{3.3}
$$

The resulting transfer function for the stage is expressed in Equation 5. The frequency responses of the nano-stage and the model are compared in Figure 3.3. Note that the frequency response of the nano-stage is obtained by taking the FFT of the experimental impulse response

Figure 3.3. The frequency response of the nano-stage (solid black) and its transfer function model G'(z) (dashed red).

3. 3. Design of the B-filter

The B filter is designed to be the inverse of the closed loop plant transfer function (see Figure 3.2), $G(z)/(1+G(z))$, based on the criteria, $R(\omega) < 1$, defined in Section 2. Instead of designing a low-pass filter, a simple static gain is used in [11]. However, the unmodeled dynamics enforce the authors of [11] to make a conservative selection for the gain value in

order to maintain the stability of the system. However, a low-pass filter ensures the stability for a wide range of frequencies. On the other hand, the inverse of $G(z)/(1+G(z))$ is noncausal since its numerator has a lower degree than its denominator. To eliminate this problem, the inverse of $G(z)/(1+G(z))$ is multiplied by another $4th$ order low pass filter having a large bandwidth. The resulting B-filter is expressed in Eq. 3.

$$
B(z) = \frac{z^8 - 2.34z^7 + 2.484z^6 - 1.615z^5 + 0.554z^4 - 0.18z^3 + 0.294z^2 - 0.261z + 0.098}{0.047z^8 + 0.045z^7 - 0.0077z^6 - 0.038z^5 - 0.012z^4}
$$
(3.5)

3. 4. Design of the Q-filter

In order to suppress the undesired harmonics at high frequencies, the bandwidth of the Qfilter must be large. As the bandwidth of the Q-filter increases, the RC is able to respond faster, but with the cost of additional noise in the control signal. For this reason, we have selected a $4th$ order model for the Q filter (our numerical simulations show that no significant improvement is obtained when a higher order model is used) and experimented with two different bandwidths: one close to the bandwidth of the stage ($\omega_q = 80$ Hz, Eq. 3.6.a) and the other is twice the bandwidth of the stage ($\omega_q = 160$ Hz, Eq. 3.6.b). The regeneration spectrum, $R(\omega)$, with $Q_1(z)$ is shown in figure 3.4.

$$
Q_1(z) = \frac{8.853 \times 10^{-5}}{z^4 - 3.612z^3 + 4.892z^2 - 2.945z + 0.6649}, \text{ (for } \omega_q = 80 \text{ Hz}) \tag{3.6.8}
$$

$$
Q_2(z) = \frac{0.0012}{z^4 - 3.262z^3 + 3.989z^2 - 2.169z + 0.4421}, \text{ (for } \omega_q = 160 \text{ Hz})
$$
\n(3.6.b)

Figure 3.4. The regeneration spectrum of the RC with Q_1 (blue dashed) and Q_2 (red solid)

3. 5. Time advances (phase compensators)

The time advances τ_q and τ_b shown in Figure 2.2 are used to compensate for the phase delays caused by the filters Q and B as well as the other delays caused by the sampling rate and the measurements. Since the transfer functions of the Q and B filters are known in advance, the initial values for these phase compensators can be estimated in advance from the phase diagrams of the filters first and then fine-tuned based on the scan performance. Since our implementation is in discrete domain, both time advances are converted to sample advances by $n = \frac{\tau}{T_s}$ where T_s is the sampling time of the system (Figure 3.2). Numerical simulations are performed in order to understand the effect of the sample advances, nq and nb, on the performance of our RC.

Chapter 4

SIMULATION MODEL

A comprehensive MATLAB/Simulink model (Figure 4.1) is developed to simulate the whole scan process and test the performance of the RC under different settings of the sample advances nq and nb. For this purpose, our original MATLAB/Simulink model [18] is further improved and modified to implement the proposed RC.

Figure 4.1. Our MATLAB/Simulink model.

The improvements include the addition of more realistic models for the nano-stage and the interactions between the scanning probe and the sample surface. In our original Simulink model [18], a simple first order model is used for the nano-stage, which is replaced by $G'(z)$ in the current study. Moreover, a standard DMT force model was utilized to estimate the nonlinear interaction forces between the AFM probe and the surface being scanned. However, in a standard DMT model, the sticking behavior of the scanning probe to the sample surface is not taken into consideration even though it is frequently observed in AFM scanning. We have further improved the DMT model to simulate the sticking effects. For this purpose, we have employed a switch, which decides if the probe sticks to the surface based on the distance between the probe's equilibrium position and the sample surface. The oscillation amplitude of the probe is reduced to zero when the probe sticks to the surface. Then, in order to separate the probe from the sample surface and hence to bring its vibration amplitude to the desired value, A_{set} , the nano-stage carrying the sample is commanded to move down along the Z-axis by the controller. As a result, the distance between the probe's equilibrium position and the sample surface increases until a threshold distance is reached, and then the probe is assumed to detach from the surface and vibrate in free air again. Here, we assume that the behavior of the probe must be almost linear for our sticking model to work well, which is valid for the stiff probe $(k = 42 \text{ N/m})$ used in our experiments. To test our assumption, we performed experiments in our AFM setup with two different probes having nominal spring constants of 42 N/m (stiff probe) and 3 N/m

(soft probe). During the experiments, the probes were actuated at their resonance frequency and the distance between the probe tip and the sample surface was slowly decreased. The results of the experiments are shown in Figure 4.2. The results validate our assumption that the probe with the lower spring constant undergoes a rapid decrease in amplitude while approaching the sample surface and a rapid increase in amplitude while retracting from the same surface. On the other hand, the stiffer probe does not exhibit such behavior. Moreover, the softer probe shows more hysteresis. Therefore, it is reasonable to conclude that the stiff probe shows a more linear behavior than a soft one and the proposed model for sticking effects can be utilized safely in the numerical simulations.

Figure 4.2. The sticking behavior of a) stiff and b) soft probes during the approaching and retracting phases.

Chapter 5

SCANNING SIMULATIONS

The scan performance of the RC was investigated using numerical simulations. Since the scan performance, in general, depends on many factors (temperature, humidity, ground vibrations, electrical and environmental noise, etc.) which are not under the full control of an experimenter, we believe that the performance of a new controller for an AFM cannot be fully appreciated without performing extensive numerical simulations.

At first, the effect of sample advances, nq and nb, on the performance of RC was investigated . The sample advances adjust the phase of the control signal in RC. Therefore, proper tuning of these parameters is critical in order to reduce the phase delays that occur during the actual scanning. For this purpose, scanning simulations were performed and the effect of sample advances on the scan performance are investigated for the range of 0 to 15 by incrementing their values by one. The goal is to obtain the optimum values returning the minimum scan error when a step profile having a width of ω_s and a height of h_S is scanned. The scan error is calculated by integrating the positional tracking error, e_X over the step width, and normalized by dividing it to the step dimensions as

$$
e_{s} = \frac{\int_{0}^{\omega_{s}} |e_{x}| dx}{\omega_{s} h_{s}}
$$
(5.1)

The iso-error curves obtained at the scan speed of 3 μ m/s are presented in Figure 5.1. One can observe from the figure that the scan error is minimum when $nq = 9$ and $nb = 6$.

Figure 5.1. Iso-error curves are generated to investigate the tracking performance of the RC by varying the sample advance parameters.

The scan performance of the RC for different values of nq and nb is shown in Figure 5.2. As shown in the figure, the sample advance nq influences the upper left and lower right corners whereas the sample advance nb influences the lower left and the upper right corners of the step.

a)

b)

Figure 5.2. The effect of a) nq when $nb = 6$ and b) nb when $nq = 9$ on the scan performance.

In Figure 5.3a, the scan performance of the RC ($nq = 9$ and $nb = 6$) is compared to that of the stand-alone PI controller at the scan speed of $3 \mu m/s$. Considering the structure given in Figure 3.2, first, the PI controller is tuned for the best performance and then the RC is turned on for the comparison. As shown in the figure, the stand-alone PI controller responds to the changes in step profile with a delay whereas the RC can respond instantly. Though the delay is small at this speed, the relative improvement in scan error is 75 % when the RC controller is used instead of the stand-alone PI controller. To compare the performance of the stand-alone PI with that of the RC at higher scan speeds, scans were also performed at the speed of 21 µm/s (see Figure 5.3b). Compared to the stand-alone PI controller, the RC reduces the scan error by 58 %. In order to achieve even better tracking results, the bandwidth of the Q filter in the RC is adjusted. When the bandwidth of the RC controller is increased (see the discussion in Section 3.4), it is observed that the relative difference in scan error is increased to 64 % (see the scan profiles for Q_1 and Q_2 in Figure 5.3b).

a)

b)

Figure 5.3. The scan results for the scan speed of a) 3 μ m/s and b) 21 μ m/s: the surface being scanned (dashed black), the results obtained by the PI controller (dash-dotted blue), the RC $(Q_1,$ solid red) and the RC with an extended bandwidth $(Q_2,$ dotted red).

One of the most important benefits of the simulation environment is that the tapping forces can be calculated from the force model directly and used as a performance measure. Since the simulations are in tapping mode, both the maximum and the deviation of instantaneous tapping forces from the nominal value are considered for the evaluation of scan results. The maximum tapping forces must be reduced to prevent damage to the probe and the sample,

especially when biological samples are scanned. The deviation of instantaneous tapping forces from the nominal value is an indicator of the scan quality. The instantaneous tapping forces as a function of lateral position are shown in Figure 5.4a and Figure 5.4b for the scan speeds of 3 μ m/s and 21 μ m/s respectively. The quantitative analysis shows that the maximum tapping force at the contact (see the instants of contact and no-contact in Figure 5.4) for the scan speed of 3 µm/s and 21 µm/s is reduced by 1% and 87% (and 95% with the extended bandwidth), respectively, when the RC is used instead of the stand-alone PI controller. Moreover, the total deviation from the nominal force value is decreased by 95% and 93% (and 98% with extended bandwidth) for the scan speeds of 3 μ m/s and 21 μ m/s, respectively.

a)

b)

Figure 5.4. The tip-sample interaction forces as a function of lateral position for the scan speeds of a) 3 μ m/s and b) 21 μ m/s under PI controller (dashed blue), RC (solid red), and RC with extended bandwidth (dotted red).

Furthermore, tracking performance of the RC to that of the stand-alone PI controller compared not only for a square wave, but also for two other wave forms: a triangular wave of amplitude 100 nm (Figure 5.5a) and a wave form made of two sinusoids having the wave lengths of 100 nm and 40 nm (Figure 5.5c). The results show that the RC outperforms the stand-alone PI controller in all cases (see the error profiles in Figures 5.5b and 5.5d).

Figure 5.5. The tracking performance of the PI (blue dashed) and RC (red solid) for a) a triangular input and c) for an input composed of two sinusoids. The corresponding error profiles are shown in b) and d) respectively.

Finally, the tracking performance and the stability of RC is further investigated under various uncertainties by running simulation scenarios. The aim in this part is to question the scan performance of the RC when something unexpected happens. Since the AFM scan process is affected by many factors most of which are not easy to detect and eliminate, we believe that the performance of any new controller should be further investigated for possible structured and unstructured uncertainties existing in the system.

Scenario 1: The proposed RC assumes that the consecutive scan lines are similar to each other and it takes a control action in the current scan line based on the error estimated for the previous scan line. However, if there is a significant difference between the consecutive scan profiles, the tracking performance of a typical RC will be poor since it can only react to this difference after one scan line is passed. In this implementation, an RC is connected in series with a standard PI controller such that the difference between the consecutive scan profiles can be compensated by the PI controller. In order to demonstrate the benefits of our approach, we design a scenario in which the step height is suddenly reduced from 100 nm to 10 nm (Figure 5.6). In this scenario, each step can be considered as one scan line for the sake of discussion. Initially, the stage tracks the 100 nm steps successfully, but when a sudden change in the step profile occurs, the repetition in the profile is broken. Then, the PI controller becomes more active and tracks the input signal until the repetition in the profile is established again. Note that the sudden reduction in the step height causes a peak in the scan profile due to the control signal carried on by the RC from the previous step is larger than the small value desired for the current step. The size of this peak can be reduced by adding a small derivative term to the PI controller. We also note that such undesired peaks do not appear at all when there is less than 50% reduction in the amplitude. On the contrary, a sudden increase in the step size does not constitute such a problem since the control signal carried on by the RC from the previous line is small, and it is dissolved in the larger control signal generated by the PI controller.

Figure 5.6. Our RC is designed as an add-on to a standard PI controller so that the PI controller can help with the tracking in case the consecutive scan profiles are significantly different from each other. In the graph above, for the sake of discussion, each step can be considered as one scan line.

Scenario 2: An AFM probe can be damaged easily during tapping-mode AFM scanning, since it is in continuous contact with the sample surface. This results in a change in the mass and consequently the resonance frequency of the probe. If undetected, the free air amplitude of the defected probe decreases, which also causes a decrease in the magnitude of the error signal transmitted to the controller. Under these circumstances, the PI controller needs to be re-tuned whereas the RC can compensate for the drop in the magnitude of the error signal and does not require re-tuning. The following simulation scenario is developed to demonstrate this concept: Initially, the probe's amplitude is set to the 70% of the free air amplitude ($A_{\text{free}} = 67 \text{ nm}$). Then, the free air amplitude is suddenly reduced to $A_{\text{free}} = 57 \text{ nm}$ without changing the set amplitude. The resulting scan profile is shown in Figure 5.7. It is evident that the PI controller suffers from the error saturation at the falling edges of the steps whereas the RC successfully tracks the surface profile.

Figure 5.7. The scan profile for a damaged probe: the actual surface (dashed black), the PI controller (dash dotted blue) and the RC (solid red).

Scenario 3: The interactions of stiff and soft probes with sample surface are investigated in section 4. As concluded, a soft probe can stick to the sample surface lot easier than a stiff one. When the soft probe sticks to the sample surface, a large error occurs and the controller generates a large control signal to move the stage downwards. This action causes undesired peaks in the scan profile. However, if the controller keeps the probe away from the sticking regime (see Figure 4.2) during the scan process, which is determined by the tip-sample separation distance, the undesired peaks can be avoided. Since the RC takes a control action based on the scan results of the previous line, the sticking behavior is handled better. In Figure 5.8, when the stand-alone PI controller is to command the stage, the soft probe sticks to the surface at the beginning of each step, causing an overshoot, which does not occur under the RC.

Figure 5.8. The scan performance of a soft probe under the PI controller (dash dotted blue) and the RC (solid red).

Chapter 6

EXPERIMENTS

Nano scanning experiments are performed with the home-made AFM setup to validate the results of the numerical simulations. A calibration grating (Micro-Masch TGZ02) consisting of steps with 3 µm pitch and 82.5 nm height $(\pm 1.5 \text{ nm})$ is scanned. The AFM probe is excited at its resonance frequency (353.8 kHz) with a free air vibration amplitude of $A_{\text{free}} = 52 \text{ nm}$ (note that the set amplitude is equal to 70% of the free air amplitude). The scanning experiments are performed at the scan speeds of $3 \mu m/s$ and $21 \mu m/s$ (Figures 6.1) and 6.2 respectively). For the scans performed under the RC, nq and nb are set to 9 and 6 respectively based on the results of the simulations presented earlier (Figure 5.1). At a first glance, the improvement achieved by the RC is not obvious for the scan speed of $3 \mu m/s$ (compare Figures 6.1a and 6.1c), but the standard PI controller follows the desired profile with a delay as observed in the simulations. The tracking error due this delay is shown in Figure 6.1b (compare it with Figure 6.1d). Since the PI controller can track the surface with a delay, the probe sticks to the rising edge of each step, causing a sudden jump in the error signal. The same phase delay also causes a sudden drop in the error signal on the falling edge of each step. As the scan speed increases to $21 \mu m/s$, the phase delay increases further

and the difference between the profiles scanned by the stand-alone PI controller and the RC becomes more obvious (Figure 6.2). Moreover, on the rising edges of the steps, the PI controller suffers from the sticking behavior of the probe, which causes a large overshoot, while on the falling edges, the PI controller suffers from error saturation, which limits its response rate. However, both problems are reduced when the RC is used. The performance of the RC becomes even better if the bandwidth of the Q filter is doubled (the last row in Figure 6.2). The results of the scanning experiments performed with our AFM set-up show that the proposed RC significantly outperforms the conventional PI controller. Compared to the stand-alone PI controller, the scan error and the average tapping forces are reduced by 66% and 58%, respectively when the scan speed is increased by 7-fold [17].

Figure 6.1. The scan profile and error signal for the scan speed of 3 μ m/s under the standalone PI controller (a, b) and the RC (c, d).

Figure 6.2. The scan profile and error signal for the scan speed of 21 μ m/s under the standalone PI controller (a, b), the RC (c, d), and the RC with extended bandwidth (e, f).

Chapter 7

DISCUSSION and CONLUSION

We investigated the use of RC in AFM scanning through scan experiments and numerical simulations. We showed that RC improves the image quality and reduces the tapping forces when compared to a stand-alone PI controller. Although the improvement in the image quality cannot be observed directly from the scan profiles at low scanning speeds, it is much clearer when the scan speed is increased. The RC is superior to the stand-alone PI controller because it is designed to compensate for the phase delays better when the input signal is repetitive. The positive effects of this phase cancellation are more obvious at high scan speeds since the problems causing a delay in the response of the stage, due to sticking of the probe to the surface and the error saturation occur more frequently. The cancellation of phase delays is possible with the help of memory loop in the RC, which utilizes the profile of the previous scan line to take a control action in the current scan line. It is also observed in the simulations that the proposed implementation of the RC as an add-on to a PI controller is effective when significant differences exist between the profiles of consecutive scan lines. However, such a large difference is unexpected in real AFM scans if sufficiently small advances are made along the y-axis.

The results of our numerical simulations also support the AFM experiments performed in our setup. Although numerical simulations have been performed in the past to investigate the cantilever dynamics and tip–sample interactions, only a few recent studies have focused on the simulation of the whole scan process [18, 19]. Since it is almost impossible to repeat a scan experiment under exactly the same conditions in real world settings due to variations in humidity, electrical noise, ground vibrations, temperature, etc, developing numerical simulations to test the performance of a new controller under different conditions is important. Moreover, even if the experimental conditions are fixed, the probe tip or the sample surface may get damaged in time, affecting the scan results adversely. All these factors make it difficult to quantitatively compare the results of experimental scans under different scan settings. On the other hand, we, for example, easily repeated the scan experiments in Simulink/Matlab environment to construct iso-curves of constant error for different values of sample advance parameters (nq and nb) of the RC to find their optimum values. In fact, those values ($nq = 9$ and $nb = 6$) also worked well for the scanning experiments performed in our AFM set-up. In addition, we readily accessed all the outputs of the simulations, some of which are not directly measurable in physical experiments (e.g. the magnitude of the tapping forces shown in Figure 12). Finally, we also designed simulation scenarios to investigate the robustness of the RC under various uncertainties. We showed that the RC does not require retuning of the parameters under the simulated uncertainties.

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