

# Dynamic Matrix Flight Control design for a nonlinear missile Autopilot

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**ABSTRACT:** Nonlinear control problem for a missile autopilot is quick adaptation and minimizing the desired acceleration to missile nonlinear model. For this several missile controllers are provided which are on the basis of nonlinear control or design of linear control for the linear missile system. In this paper a linear control of dynamic matrix type is proposed for the linear model of missile. In the first section, an approximate two degrees of freedom missile model, known as Horton model, is introduced. Then, the nonlinear model is converted into observable and controllable model base on the feedback linear rule of input-state mode type. Finally for design of control model, the dynamic matrix flight control, which is one of the linear predictive control design methods on the basis of system step response information, is used. This controller is a recursive method which calculates the development of system input by definition and optimization of a cost function and using system dynamic matrix. So based on the applied inputs and previous output information, the missile acceleration would be calculated. Unlike other controllers, this controller doesn't require an interaction effect and accurate model. Although, it has predicting and controlling horizon, there isn't such horizons in non-predictive methods.

*Keywords:* Dynamic Matrix Control (DMC), 2 degrees of freedom missiles model, Feedback linearization, Flight control

# INTRODUCTION

The performance of aerospace systems such as aircraft, spacecraft and missiles is highly dependent on the capabilities of the guidance, navigation and control systems. To achieve improved performance in such aerospace systems, it is important that more sophisticated control systems be developed and implemented (Tsourdos & White, 2005). Modern missiles often operate in flight regimes where nonlinearities significantly affect dynamic response. For example, a high performance missile must be quickly responsive to and follow accurately any guidance commands, so that, it can intercept fast moving and agile targets (Bryson & Y-C 1975; Xin & Balakrishnan, 2003). The tracking performance of a missile is also dependent on the location within the flight envelope and varies with factors such as Mach number and dynamic pressure. Many nonlinear control methods have been proposed for the missile autopilot design. Several approaches, including adaptive control (Lin & Cloutier, 1991), nonlinear control (White et al., 1998), and gain scheduling (Shamma & Cloutier, 1993) have been used to alleviate these tracking problems.

In this paper, input-state linearization method is used for linearization. Feedback linearization is a popular method used in nonlinear control applications, and there have been several flight control demonstrations (Snell, 1992). Feedback linearization methods can be viewed as the ways of algebraically transforming a nonlinear system dynamic, fully or partially, into a simple linear one. In the standard approach to exact input-state linearization, one uses coordinate transformation and static state feedback such that the closed-loop system, in the defined region, takes a linear canonical form (Cheh et al., 2004). The input-state linearization method differs from conventional linearization techniques in that input-state linearization is achieved via exact state transformations and feedback, rather than by linear approximations of the system dynamics. Linearization has a number of advantages in the control field, including simplicity, being easily implemented, being well developed, and so forth, and thus the literature contains many studies of the linearization form is obtained, the linear control design scheme is employed to achieve stabilization or tracking (Slotine, 1991).

Model predictive control designates a wide range of control methods which make an explicit use of a model of the process to obtain the control signal by minimizing an objective function. The common idea is

receding strategy which means that, the objective function is minimized by considering also future control actions along the so-called prediction horizon but only the first control signal is applied to the system; then the horizon is displaced towards the future and the next control signal is recalculated. In view of the main idea of

the model predictive control strategy it is clear that, the process model plays an important role in the method. There are many types of models used; one of them is the so-called step response model. For stable systems one may take a truncated response and with this model the control signal can be calculated by minimizing the objective function along the prediction horizon. The step response model is used by the Dynamic Matrix Control (DMC) algorithm, developed at the end of the seventies (Cutler & Ramaker, 1979 and 1980). It is a successful and widely applying technique in industrial and engineering applications (Besnyei & Simon, 2010). The dynamic matrix control is one of the important and representative predictive control algorithms and characterized by replacing the traditional auto correction of single step prediction with multiple step prediction, repeating optimization based on the practical feedback information, and restraining the algorithm sensitivity to parameter change of the model effectively (LU & TSA, 2007; Yao & Guo, 2006; Guo et al., 2010).

The purpose of this paper is to find a control rule for a nonlinear missile, so that, by applying it, the desired acceleration of the missile be obtainable. Several controllers for this nonlinear missile model have been proposed which present a combination of feedback linearization methods or local linearization with linear, adaptive and intelligent controllers.

For this model, the adaptive controller by varying parameters of the missile condition (Tsourdos & White, 2005), multi-objective fuzzy controller (Tsourdos et al., 2006), variable gain fuzzy controller (Tabataba et al., 2010) and quasi-LPV controller are designed, which these controllers are designed using the input - output feedback linearization or local linearization (White et al., 2007). This paper presents a dynamic matrix controller which is the controller for model missile.

Due to the predictive and controllable horizons, which aren't expressed in unpredictable controllers, this algorithm is not comparable with unpredictable methods. However, due to its simplicity and good performance of the algorithm compared to other predictive algorithms, designers apply this algorithm to design the controller.

For example, the model algorithm control (MAC), which is based on the impulse response, has two main problems, one of them is lack of accurate impulse response in operational systems, and the other is the designed control signal due to lack of control effort in inaccurate cost function.

Another method of widely used predictive method is the Generalized Predictive Control (GPC), which is on the basis of the discrete transfer function model of the system. This method is generalized method of two previous ones, which get better results than the previous methods, but heavy calculations of this method prevents from simple design for complex systems.

In the first part of this paper, the nonlinear models and design of linearization rule and linearization method of input - state are discussed.

In this linearization, one of the problems of designing a linearization rule which is changes of the aerodynamic coefficients based on the parameters such as Mach number and angle of attack is resolved. Furthermore, the dynamic matrix control for step response of linear system which has limited final value is designed.

Simulation results show that, the designed linear rule was succeeded and the applied dynamic matrix control has a suitable and robust adaptation for the missile acceleration.

# Missile model

The missile model used in this study derives from a nonlinear model produced by Horton of Matra-British Aerospace (Horton, 1992). This study will look at the reduced problem of a 2 DOF controller for the pitch and yaw planes without roll coupling. This model describes a reasonably realistic airframe of a tail controlled tactical missile in the cruciform fin configuration (Fig.1). The angular and translational equations of motion of the missile airframe are given by

$$\dot{v} = \frac{1}{2m} \rho V_0 S \left( C_{yv} v + V_0 C_{y\xi} \xi \right) - Ur$$
  
$$\dot{r} = \frac{1}{2} I_z^{-1} \rho V_0 S d \left( \frac{1}{2} d C_{nr} r + C_{nv} v + V_0 C_{n\xi} \xi \right)$$
(1)

where the variables are defined in Fig. 1, Table 1 and 2.



Figure 1. Airframe axes.

In Eq. (1), v is the sideslip velocity, U is the longitudinal velocity, r is the body rate,  $\zeta$  is the rudder fin deflections, V<sub>0</sub> is total velocity vector and calculated from  $\sqrt{U^2 + v^2}$  and V<sub>0</sub>≈U because U>>v, C<sub>yv</sub>, C<sub>yζ</sub> is lateral force coefficient due to sideslip velocity and fin angle, C<sub>nr</sub>, C<sub>nv</sub>, C<sub>nζ</sub> is yawing moment coefficient due to body rate, sideslip velocity and fin angle. These forces and moments are derived from wind tunnel measurements and by using polynomial approximation algorithms which can be fitted to the set of curves taken from look-up tables for different flight conditions (Tsourdos & White, 2000). A detailed description of this model can be found in Horton (1992) found that some of them are shown in Table 1.

Table 1.		
Force and moment coefficients of missile model		
$C_{yv}$	$0.5\Big[\big(-25+M-60\big \sigma\big \big)\big(1+\cos 4\lambda\big)$	
	$+ (-26 + 1.5M - 30 \sigma )(1 - \cos 4\lambda) ]$	
$C_{y\xi}$	$10 + 0.5 \Big[ \Big( -1.6M + 2  \sigma  \Big) \Big( 1 + \cos 4\lambda \Big)$	
	$+(-1.4M + 1.5 \sigma )(1-\cos 4\lambda)$	
$C_{nr}$	$-500 - 30M + 200  \sigma $	
$C_{nv}$	$S_m C_{yy}$ , where	
	$S_m = d^{-1} \left[ 1.3 + 0.1M + 0.2 (1 + \cos 4\lambda) \right] \sigma$	
	$+0.3(1-\cos 4\lambda) \sigma -(1.3+m/500)]$	
$C_{n\xi}$	$S_f C_{y\xi}$ , where	
	$S_f = d^{-1} \left[ 2.6 - (1.3 + \frac{m}{500}) \right]$	

These force and moment coefficients depending on parameters such as  $\lambda$ , Mach number and  $\sigma$ .  $\lambda$  is roll angle, The Mach number is defined as M=V<sub>0</sub>/a where a is the speed of sound.  $\sigma$  is incidence angle or the angle between  $\vec{v}$  and  $\vec{V_0}$ , can be taken as  $\sigma$ =v/V<sub>0</sub>, as sin $\sigma$ ≈ $\sigma$  for small  $\sigma$  (Tsourdos et al. 2006). Involving the Mach number  $M \in [0.6 \ 6]$ , roll angle  $\lambda \in [4.5^{\circ} \ 45^{\circ}]$  and total incidence  $\sigma \in [3^{\circ} \ 30^{\circ}]$  (White et al. 2006). In Table 2 the characteristics of the Horton model are presented.

Table 2.			
Characteristics of the Horton model			
$ ho_{0}$	Sea level air density	$1.23 Kg/m^3$	
ρ	Air density	$\rho = \rho_0 - 0.094h$	
h	Altitude	in Km	
$a_0$	Sea level velocity of sound	340 <i>m/s</i>	
а	Velocity of sound	$a = a_0 - 4h m/s$	
d	Reference diameter (calibre)	0.2 <i>m</i>	
S	Reference area	$\pi d^2/4 = 0.0314m^2$	
l	Length of the missile	2.7 <i>m</i>	
т	Mass	125Kg	
$I_z$	Lateral inertia	$75 Kgm^2$	
$S_m$	Static margin (body+wings)	$x_{cp} - x_{cg}$	
$S_{f}$	Fin moment arm for lateral motion	$x_f - x_{cg}$	
$x_{f}$	Centre of pressure (fins only)	2.6 m	
$X_{cp}$	Centre of pressure (body+wings)	$x_{cp} = 1.3 + 0.1M + 0.2  \sigma $	
$x_{cg}$	Centre of gravity	1.55m	

# Input – State linear feedback

The standard form of the nonlinear system can be written as

$$\dot{x} = f(x) + g(x)\zeta$$

(2)

where f(x) and g(x) are smooth vector fields on  $R^n$ . If the state vector is defined as x=[v r]<sup>T</sup>, then Eq. (1) can be rewritten in the standard form as in Eq. (2) with the vector fields defined as follows:

$$f(x) = \begin{bmatrix} \frac{1}{2m} \rho V_0 S C_{yv} v - Ur \\ \frac{1}{2} I_y^{-1} \rho V_0 S d(\frac{1}{2} d C_{mr} r + C_{nv} v) \end{bmatrix}$$

$$g(x) = \begin{bmatrix} \frac{1}{2m} \rho V_0^2 S C_{y\xi} \\ \frac{1}{2} I_y^{-1} \rho V_0^2 S d C_{n\xi} \end{bmatrix}$$
(3)

Definition 1. A nonlinear system in the form of Eq.(2) is said to be input-state linearizable if there is a region  $\Omega$  in R<sup>n</sup>, a diffeomorphism  $\Phi: \Omega \rightarrow R^n$ , and a nonlinear feedback control law and defining a new input as (Slotine, 1991):

$$\xi = \alpha(v, r) + \beta(v, r)u$$

(5)

Such that the new state variables  $Z = \Phi(x)$  and the new input u satisfy a linear time-invariant relation:

$$\dot{Z} = AZ + Bu$$

where u is new input or linear model input, the new state Z is called the linearizing state, and the control law (4) is called the linearizing control law. By the above definition, the fundamental results of feedback linearization can be used to linearize the system.

Select the matrix A and B as

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \qquad B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$
(6)

Linear system is controllable and observable. Expanding Eq. (5), we obtain

$$\frac{\partial z_1}{\partial x} f(x) = z_2 \quad \frac{\partial z_1}{\partial x} g(x) = 0$$

$$\frac{\partial z_2}{\partial x} f(x) = 0 \quad \frac{\partial z_2}{\partial x} g(x) = u$$
(7)

Since  $z_1,\,z_2$  are independent of  $\xi,$  so  $\alpha(v,r),\,\beta(v,r)$  is calculated as

$$\alpha = -\frac{\begin{bmatrix} \frac{\partial z_2}{\partial v} & \frac{\partial z_2}{\partial r} \end{bmatrix} \begin{bmatrix} F_1 \\ F_2 \end{bmatrix}}{\begin{bmatrix} \frac{\partial z_1}{\partial v} & \frac{\partial z_1}{\partial r} \end{bmatrix} \begin{bmatrix} G_1 \\ G_2 \end{bmatrix}} = -\frac{\frac{\partial z_2}{\partial v}F_1 + \frac{\partial z_2}{\partial r}F_2}{\frac{\partial z_2}{\partial v}G_1 + \frac{\partial z_2}{\partial r}G_2}$$

$$\beta = \frac{1}{\begin{bmatrix} \frac{\partial z_1}{\partial v} & \frac{\partial z_1}{\partial r} \end{bmatrix} \begin{bmatrix} G_1 \\ G_2 \end{bmatrix}} = \frac{1}{\frac{\partial z_2}{\partial v}G_1 + \frac{\partial z_2}{\partial r}G_2}$$
(8)

where

$$F_{1}(C_{yy}, V_{0}, v, r) = \frac{1}{2m} \rho V_{0} S C_{yy} v - Ur$$

$$F_{2}(C_{nv}, C_{nr}, V_{0}, v, r) = \frac{1}{2} I_{y}^{-1} \rho V_{0} S d \left(\frac{1}{2} d C_{nr} r + C_{nv} v\right)$$

$$G_{1}(C_{y\xi}, V_{0}) = \frac{1}{2m} \rho V_{0}^{2} S C_{y\xi}$$

$$G_{2}(C_{n\xi}, V_{0}) = \frac{1}{2} I_{y}^{-1} \rho V_{0}^{2} S d C_{n\xi}$$
(9)

Assuming the  $z_1=v$  then

$$z_2 = F_1 - \frac{G_1}{G_2} F_2 \tag{10}$$

With  $z_2$  be calculated the following relationships.

$$\frac{\partial z_2}{\partial v} = \frac{\partial F_1}{\partial v} - \frac{\left(\frac{\partial G_1}{\partial v}F_2 + \frac{\partial F_2}{\partial v}G_1\right)G_2 - G_1F_2\left(\frac{\partial G_2}{\partial v}\right)}{G_2^2}$$

$$\frac{\partial z_2}{\partial r} = \frac{\partial F_1}{\partial r} - \frac{\left(\frac{\partial F_2}{\partial r}G_1\right)G_2}{G_2^2} = \frac{\partial F_1}{\partial r} - \frac{G_1}{G_2}\frac{\partial F_2}{\partial r}$$
(11)

It should be noted that  $C_{yv}$ ,  $C_{y\zeta}$ ,  $C_{nr}$ ,  $C_{nv}$ ,  $C_{n\zeta}$  are related to M,  $\sigma$  that M,  $\sigma$  is related to v. Some derivatives as

$$\frac{\partial V_0}{\partial v} = \frac{v}{\sqrt{U^2 + v^2}} = \sigma$$

$$\frac{\partial M}{\partial v} = \frac{1}{a} \frac{v}{\sqrt{U^2 + v^2}} = \frac{\sigma}{a}$$

$$\frac{\partial \sigma}{\partial v} = \frac{\sqrt{U^2 + v^2} - \frac{v}{\sqrt{U^2 + v^2}}}{(U^2 + v^2)} = \frac{U^2}{V_0^3} \qquad (12)$$

$$\frac{\partial |\sigma|}{\partial v} = \begin{cases} \frac{U^2}{V_0^3} & v > 0\\ -\frac{U^2}{V_0^3} & v < 0 \end{cases}$$

If force and moment coefficients is defined as

$$C_{yv} = C_{yv0} + C_{yvM} M - C_{yv|\sigma|} |\sigma|$$

$$C_{y\xi} = 10 - C_{y\xi M} M + C_{y\xi|\sigma|} |\sigma|$$

$$C_{nr} = -500 - 30M + 200 |\sigma|$$

$$S_{m} = S_{m0} + S_{mM} M + S_{m|\sigma|} |\sigma|$$

$$C_{nv} = S_{m} C_{yv}$$

$$C_{n\xi} = S_{f} C_{y\xi}$$
(13)

Then derivatives are as follows.

$$\frac{\partial C_{yv}}{\partial v} = C_{yvM} \frac{\sigma}{a} - C_{yv|\sigma|} \frac{\partial |\sigma|}{\partial v}$$

$$\frac{\partial C_{y\xi}}{\partial v} = -C_{y\xiM} \frac{\sigma}{a} + C_{y\xi|\sigma|} \frac{\partial |\sigma|}{\partial v}$$

$$\frac{\partial C_{nr}}{\partial v} = -30 \frac{\sigma}{a} + 200 \frac{\partial |\sigma|}{\partial v}$$

$$\frac{\partial S_m}{\partial v} = S_{mM} \frac{\sigma}{a} + S_{m|\sigma|} \frac{\partial |\sigma|}{\partial v}$$

$$\frac{\partial C_{nv}}{\partial v} = \frac{\partial S_m}{\partial v} C_{yv} + \frac{\partial C_{yv}}{\partial v} S_m$$

$$\frac{\partial C_{n\xi}}{\partial v} = S_f \left( -C_{y\xiM} \frac{\sigma}{a} + C_{y\xi|\sigma|} \frac{\partial |\sigma|}{\partial v} \right)$$
(14)

Using the Eqs (8) to (15)  $\alpha$ ,  $\beta$  can be calculated. Missile acceleration is second state of linear model and it is calculated as

$$y = CZ$$

$$C = \begin{bmatrix} 0 & 1 \end{bmatrix}$$
(15)

In this linearization, due to taking into account the effect of aerodynamic coefficients in the equations of the linear rule, the effect of these parameters on the model output can be neglected.

In a previous study (Tsourdos & White, 2005), the designed linear rule of input - output has been calculated with the assumption of constant aerodynamic coefficients and regardless of the effect of missile conditions on these parameters. This proposed control rule solves one of the main problems of missile controlling which is the change of aerodynamic coefficients. The other effective parameters in this model such as changes in mass, air pressure, Inertia, etc. can be removed by a time-varying gain or a smart filter.

#### Step Responses

For designing DMC controller, step response is required. Linear system transfer function is as follows:

$$\frac{y(s)}{u(s)} = \frac{1}{s} \tag{16}$$

According to Eq. (16), the output step response of the linear system seen in Fig 2.



Figure 2. Step responses for linear missile acceleration

It is difficult to design DMC controller, because transfer function has poles on the  $j\omega$  axis. To remove ramp of the step responses, the feedback is used as follows.

$$\dot{Z} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} Z + B \left( u - \begin{bmatrix} 0 & k_2 \end{bmatrix} Z \right)$$
(17)

Assuming  $k_2=5$ , then step responses figures is presented in Fig 3.



Figure 3. Step responses with state feedback for linear model

# Dynamic Matrix Flight Control

DMC is a control based optimization methodology that is explicitly utilized in dynamic mathematical model of a process to obtain a signal which minimizes the objective function. The advantages of this controller include: controlling sensitivity to time delay, interaction with other states on output and good resistance to noise input due to the recursive algorithm which is the superiority of this controller compared to other linear controllers in linear processes. This model which uses the step response of a system to design a cost function, describes the system perfectly (Paulusová & Dúbravská 2010, Bemporad & Morari 1999, Carlos et al. 1989).

The model must describe the system well. The future process outputs y(k+i) for i=1,....,p are predicted over the prediction horizon (p) using a model of the process. These values depend on the current process state and on the future control signals u(k+i) for i=0,....,m-1 over the control horizon (m), where m≤p. The control variable is manipulated only within the control horizon and remains constant afterwards, u(k+i)=u(k+m-1) for i=m,....,p-1. Process interactions and dead times can be intrinsically handled with model predictive control schemes such as DMC (Paulusová & Dúbravská 2010, Bemporad & Morari 1999, Carlos et al. 1989). The principle of DMC is shown in Fig. 4.

A general objective function is the following quadratic form

$$j = (y_d - y_p)^T Q (y_d - y_p) + \Delta U^T R \Delta U$$
(18)
  
past future
  
Predicted outputs
  
Manipulated u(t+k)
  
Inputs
  
t+n
  
t+p
  
(18)



t + 1 + m

t+'1

(19)

t+1 t+2

where,  $y_d$  is desired set point, R and Q are weight identity matrix, p and m are length of the prediction horizon and control horizon,  $\Delta u(k)$  is change in manipulation variable and calculated as

$$\Delta u(k) = u(k) - u(k-1)$$

 $y_{p}$  is the process output, at sample instant is given as

$$y_{p}(k) = \sum_{i=1}^{\infty} g_{i} \Delta u(k-1)$$
 (20)

where g<sub>i</sub> is step response coefficients.

By minimizing objective function, the optimal solution is then given in matrix form as:

$$\Delta U = \left(G^T Q G + R\right)^{-1} G^T Q E \tag{21}$$

where G is dynamic matrix and constructed as

$$G = \begin{bmatrix} g_1 & 0 & \dots & 0 \\ g_2 & g_1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ g_m & g_{m-1} & \dots & g_1 \\ \vdots & \vdots & \ddots & \vdots \\ g_p & g_{p-1} & \dots & g_{p-m+1} \end{bmatrix}$$
(22)

and E is defined as

$$E = y_d - y_{past} - D \tag{23}$$

In Eq (23), D is disturbances are considered to be constant between sample instants which can be removed by filter as

(25)

$$f(z) = \frac{1 - \gamma}{1 - \gamma z^{-1}}$$
(24)

where  $\gamma$  is between 0 and 1.  $y_{past}$  in Eq (23) is calculated as

$$y_{past} = G^{-1} \Delta U^{-} + U_{DMC} g_{N+1}$$

where,  $g_{N+1}$  is final value of step response and G<sup>-</sup> constructed as



Figure 5. Proposed flight control scheme for tracking based on DMC.

$$G^{-} = \begin{bmatrix} g_{2} & g_{3} & \cdots & \cdots & g_{N-1} & g_{N} \\ g_{3} & g_{4} & \cdots & \cdots & g_{N} & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ g_{p+m} & \cdots & g_{N} & 0 & \cdots & 0 \end{bmatrix}$$
(26)

In the previous equations, matrix elements of the  $\Delta U$ ,  $\Delta U^{-}$  and U as

$$\Delta U = \begin{bmatrix} \Delta u(k) & \Delta u(k+1) & \dots & \Delta u(k+m-1) \end{bmatrix}^{T}$$

$$\Delta U^{-} = \begin{bmatrix} \Delta u(k-1) & \Delta u(k-2) & \dots & \Delta u(k-N+1) \end{bmatrix}^{T}$$

$$U_{DMC} = \begin{bmatrix} u(k-N) & u(k-N+1) & \dots & \Delta u(k-N+p-1) \end{bmatrix}^{T}$$
(27)

Optimal input in present time as

$$u(k) = u(k-1) + \Delta u(k)$$
(28)

# SIMULATION RESULTS

For simulation, the desired output applied in the following equation:

$$\int_{0}^{\infty} y_{d} dt \le y_{d \max}$$
(29)

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where  $y_{dmax}$  is a maximum value for  $y_d$ .

This relationship must be applied to model assumptions. Parameters for DMC are assumed as:

$$p = 25$$
  $m = 5$  (30  
 $R = 0.5I_{p \times p}$   $Q = 0.5I_{m \times m}$  )

The sample time T=0.04, and the truncated horizon N=42. To test the performance of the controllers, simulations under desired output are performed.

To simulate the performance of the controller, 4 desired output is considered for it. In cases of 1 and 2, a periodic signal and a pulse and in the case of 3, an exponential signal is applied to the system as the desired inputs. In case 4 also, the signal in case 1 plus a white noise with zero mean and variance of 1 is assumed as input.



Figure 6. Time responses of the controlled and reference variable under DMC for Case 1



Figure 7. Time responses of the controlled and reference variable under DMC for Case 2



Figure 8. Time responses of the controlled and reference variable under DMC for Case 3



Figure 9. States of nonlinear and linear models with rudder and linear model input for Case 1



Figure 10. Time responses of the controlled and reference variable under DMC for Case 4

# CONCLUSION

In this paper, for the first time the idea of DMC controllers for flight control of a missile which has nonlinear performance is implemented. In this controller, the step response model is defined as a parametric model and it is used for the design and optimization of the cost function and optimal selection of inputs. The advantages of this method are its easy implementation compared with the adaptive methods and lack of sensitivity to the interaction and its performance for the acceleration response of the missile.

Computer simulation results of the missile show that for intercept of the target point, the proposed method has a good performance. The resistance to noise of the proposed method under these constraints is quite visible, although it has a poor function in ramp response, but this problem is solved with a feedback mode model. Due to incorporate the aerodynamic coefficients as a function of lateral velocity of missile, the designed control law has proper and accurate performance which observed in Fig. 6 & 9.

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