1 Theoritical study

For tracking, the point s is used. Point s is the origin of the steerable wheel frame(R_s,refer Figure 1.1). The orientation of this frame with respect to the R_0 frame is $\theta+\beta$. A new variable ψ is defined, where $\psi = \theta+\beta$. In the frame R_0, the posture to be tracked is:

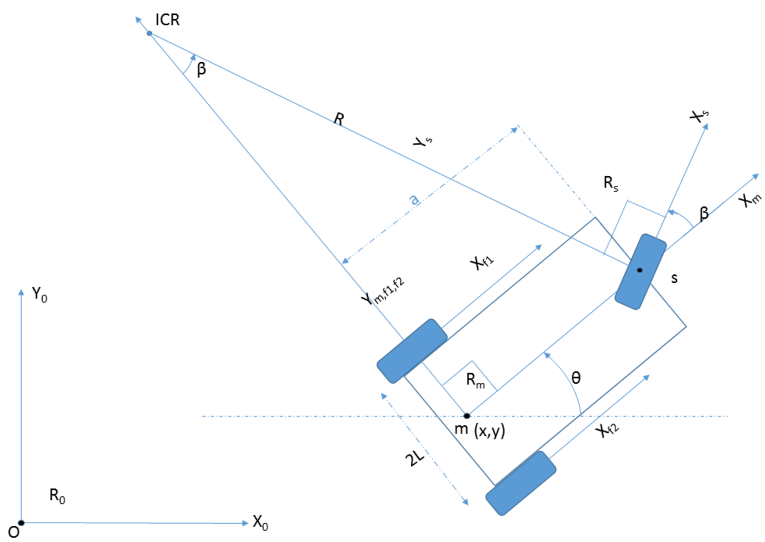

Figure 1.1: Schematic of Type 11 Robot

$$
\begin{aligned}
x_s &= x + a\cos\theta \\
y_s &= y + a\sin\theta \\
\psi &= \theta + \beta
\end{aligned}
$$

Here, (x,y) is the coordinate of the point m in the R_0 frame. On taking the time derivative of the equations above, the following equation are obtained:

$$
\begin{aligned}
\dot{x}_s &= \dot{x} - a\dot{\theta}\sin\theta \\
\dot{y}_s &= \dot{y} + a\dot{\theta}\cos\theta \\
\dot{\psi} &= \dot{\theta} + \dot{\beta}
\end{aligned}
$$

Using the the posture kinematic model of section 2, the above equations can be written as:

$$\begin{aligned}
\dot{x}_s &= a\cos\theta\cos\beta u_m - a\sin\theta\sin\beta u_m = a\cos(\theta+\beta)u_m = a\cos\psi u_m \\
\dot{y}_s &= a\sin\theta\cos\beta u_m + a\cos\theta\sin\beta u_m = a\sin(\theta+\beta)u_m = a\sin\psi u_m \\
\dot{\psi} &= \sin\beta u_m + u_s
\end{aligned}$$

$\implies$

$$\begin{bmatrix} {}^0\dot{x}_s \\ {}^0\dot{y}_s \\ \dot{\psi} \end{bmatrix} = \begin{bmatrix} a\cos\psi & 0 \\ a\sin\psi & 0 \\ \sin\beta & 1 \end{bmatrix} \begin{bmatrix} u_m \\ u_s \end{bmatrix}$$

This is the rate of change of state variables of the actual robot. Similarly, the virtual robot (the reference), which needs to be tracked (or followed) can be described by the following equations:

$$\begin{bmatrix} {}^0\dot{x}_{sr} \\ {}^0\dot{y}_{sr} \\ \dot{\psi}_r \end{bmatrix} = \begin{bmatrix} a\cos\psi_r & 0 \\ a\sin\psi_r & 0 \\ \sin\beta_r & 1 \end{bmatrix} \begin{bmatrix} u_{mr} \\ u_{sr} \end{bmatrix}$$

All the state space equations developed above are with respect to the frame R_0.
Let error be defined as: $error = reference - actual$. In the R_s frame, the errors in rate of change of x,y and ψ is:

$$\begin{bmatrix} {}^s\dot{x}_e \\ {}^s\dot{y}_e \\ \dot{\psi}_e \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ -\dot{\psi} \end{bmatrix} \times \begin{bmatrix} {}^s x_e \\ {}^s y_e \\ \psi \end{bmatrix} +^s \Omega_0(\psi) \left[\begin{bmatrix} {}^0\dot{x}_{sr} \\ {}^0\dot{y}_{sr} \\ \dot{\psi}_r \end{bmatrix} - \begin{bmatrix} {}^0\dot{x}_s \\ {}^0\dot{y}_s \\ \dot{\psi} \end{bmatrix} \right]$$

Here,

$$\Omega(\psi) = \begin{bmatrix} \cos\psi & \sin\psi & 0 \\ -\sin\psi & \cos\psi & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Using the equations developed in this section, the error model is simplified as:

$$\begin{bmatrix} {}^s\dot{x}_e \\ {}^s\dot{y}_e \\ \dot{\psi}_e \end{bmatrix} = \begin{bmatrix} a u_{mr}\cos\psi_e \\ a u_{mr}\sin\psi_e \\ u_{mr}\sin\beta_r + u_{sr} \end{bmatrix} + \begin{bmatrix} {}^s y_e \sin\beta - a & {}^s y_e \\ -{}^s x_e \sin\beta & -{}^s x_e \\ -\sin\beta & -1 \end{bmatrix} \begin{bmatrix} u_m \\ u_s \end{bmatrix}$$

In short, the error model can be written as: $\dot{X} = f(X, \mathbf{u})$ Here, X is the state variable vector and $\mathbf{u}$ is the control input vector. Consider a Lyapunov function $V(X)$ for the system $\dot{X} = f(X, \mathbf{u})$. Designing a Lyapunov controller means finding a control input vector $\mathbf{u}$ to satisfy:

$$\dot{V}(X) = \frac{\partial V}{\partial X}\dot{X} = \frac{\partial V}{\partial X} f(X, \mathbf{u}) < 0$$

The Lyapunov function chosen must be positive definite. For this system, the Lyapunov function chosen is:

$$V(X) = \frac{1}{2}({}^s x_e^2 +^s y_e^2 + \frac{{}^s\psi_e^2}{K_y})$$

It can be guaranteed that this is a Lyapunov function because it is positive definite and clearly has continous partial derivatives.

$$\dot{V}(X) = \frac{\partial V}{\partial X}\dot{X} = \frac{\partial V}{\partial X} f(X, \mathbf{u}) < 0$$

$\implies$

$$\begin{bmatrix} {}^s x_e & {}^s y_e & \frac{\psi_e}{K_y} \end{bmatrix} \begin{bmatrix} {}^s\dot{x}_e \\ {}^s\dot{y}_e \\ \dot{\psi}_e \end{bmatrix} < 0$$

On simplifying:

$$(au_{mr}\cos\psi_e - au_{mr})^s x_e + (\frac{au_{mr}\sin\psi_e^s y_e}{\psi_e} + \frac{u_{mr}\sin\beta_r + u_{sr}}{K_y} - \frac{u_{mr}\sin\beta + u_{sr}}{K_y})\psi_e < 0$$

This inequality is satisfied if:

$$au_{mr}\cos\psi_e - au_{mr} = -K_x^s x_e \tag{1}$$

$$\frac{au_{mr}\sin\psi_e^s y_e}{\psi_e} + \frac{u_{mr}\sin\beta_r + u_{sr}}{K_y} - \frac{u_{mr}\sin\beta + u_{sr}}{K_y} = -\frac{K_\psi}{K_y}^s \psi_e \tag{2}$$

$\implies$

$$u_m = \frac{au_{mr}\cos\psi_e + K_x^s x_e}{a} \tag{3}$$

$$u_s = K_y\frac{au_{mr}\sin\psi_e^s y_e}{\psi_e} + K_\psi\psi_e + u_{mr}\sin\beta_r + u_{sr} - u_{mr}\sin\beta \tag{4}$$

The above equations yield the value of the control inputs.
Hence,

$$\mathbf{u} = \begin{bmatrix} \frac{au_{mr}\cos\psi_e + K_x^s x_e}{a} \\ \frac{K_y au_{mr}\sin\psi_e^s y_e}{\psi_e} + K_\psi\psi_e + u_{mr}\sin\beta_r + u_{sr} - u_{mr}\sin\beta \end{bmatrix}$$

The use of this control law makes the system globally asymptotically stable because the substitution of these values in $\dot{V}(x)$ yields:

$$\dot{V}(x) = -K_x^s x_e^2 - K_\psi\frac{\psi_e^2}{K_y}$$

Choice of positive gains ensures $\dot{V} < 0$ globally, hence the stability.
As can be seen in the equations 3 and 4, the knowledge of reference steering angle β_r, the reference steering angular velocity u_s and the reference translational velocity of the steering wheel u_m is required.
The control objective is to force the robot to move along a circular trajectory centred at $(0,0)$ with radius $R = 2$ m and angular velocity of $\omega_d = 0.5$ rad/s, this implies that the tangential velocity of the robot must be 1 m/s.
Since the robot is expected to follow a circular trajectory, the reference steering angle must be a constant. From Figure 1.1 it can be concluded that for a circular path, the constant reference steering angle is:

$$\beta_r = \sin^{-1}\frac{a}{R}$$

The translational velocity of the type (1,1) robot is $\dot{x}^2 + \dot{y}^2 = au_m cos\beta$ must be equal to the tangential velocity required to satisfy the control objective. If the tangential velocity is $V_{tangential}$ then:

$$\dot{x}^2 + \dot{y}^2 = au_m cos\beta = V_{tangential}$$

$\implies$

$$u_m = \frac{V_{tangential}}{a\cos\beta}$$

For the control objective specified, $V_{tangential} = R\omega_d = 1$ m/s.

2 Modelling

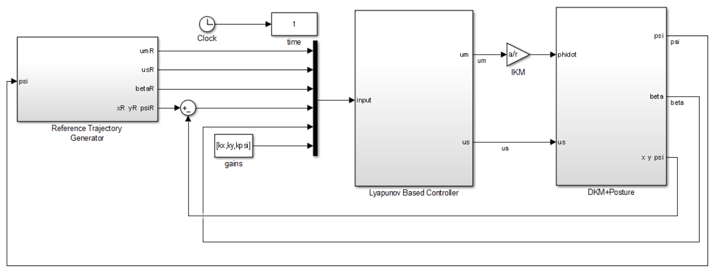

Figure 2.1: Simulink model of a Lyapunov function controlled Type(1,1) robot

Figure 5.1 presents the `simulink` model of the Lyapunov function controlled Type(1,1) robot. A brief description of all the subsystems is presented below:

- The `Reference Trajectory Generator` subsystem generates the desired trajectory, in the R_s frame, that the robot is expected to follow. Besides these, it also generates the reference steering wheel translational velocity u_{mr}, the reference steering wheel angular velocity u_{sr} and the reference steering angle β_r.
- The `Lyapunov Based Controller` subsystem implements the Lyapunov control law deduced in the previous section. It takes the errors in position (expressed in R_s frame) & orientations , u_{mr}, u_{sr}, β_r, β and the controller gains as inputs and gives $\mathbf{u} = [u_m, u_s]^T$ as the output.
- The `DKM+Posture` block takes as input the steering wheel spin velocity $\dot{\phi}$ and the the steering angular velocity u_s as input. The subsytem implements the direct kinematic model and the posture kinematic model for the type (1,1) robot. There is a `Localisation` block inside this subsytem which shifts the tracking point to point s. Then a frame transformation is performed and the subsytem outputs the posture of the robot in the R_s frame.

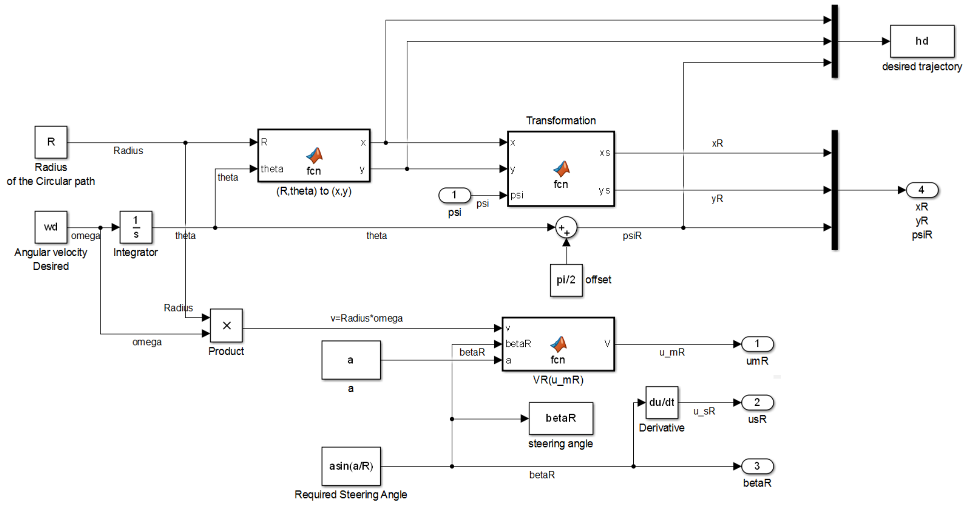

Figure 2.2: Reference Trajectory Generator subsystem

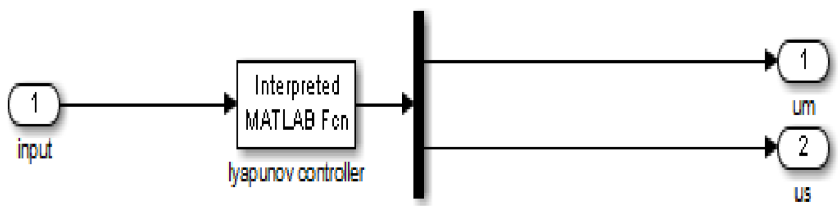

Figure 2.3: Lyapunov Based Controller subsystem

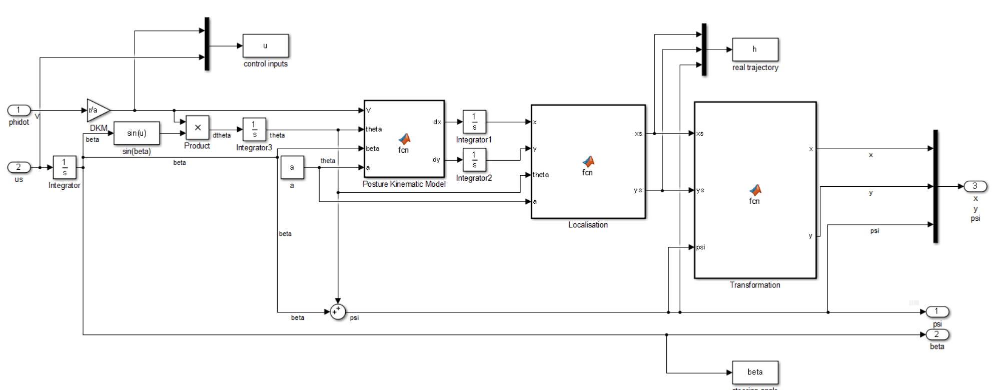

Figure 2.4: DKM+Posture subsystem

2.1 Validation and Simulation

Remark: The `fixed step ode4 Runge Kutta` solver is used with a time step of 0.001 second.
Tuning: K_x,K_y and K_θ
To ensure $\dot{V} < 0$, K_x,K_y and K_θ must all be positive.

- Tuning with $K_x = 1$, $K_y = 1$ and $K_\theta = 1$:

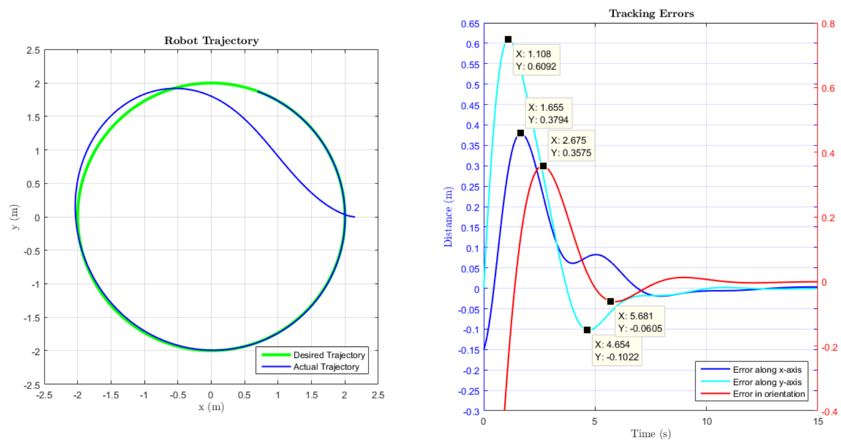

Figure 2.5: Robot Trajectory and errors in R_0 frame with $K_x = 1$, $K_y = 1$ and $K_\theta = 1$

It can be clearly noted in the Figure 5.5 that the trajectory is stable and converges close to the reference, but the response time is high and the gains need to be re tuned to give better dynamic performance. Increasing the gains may improve the dynamic performance.

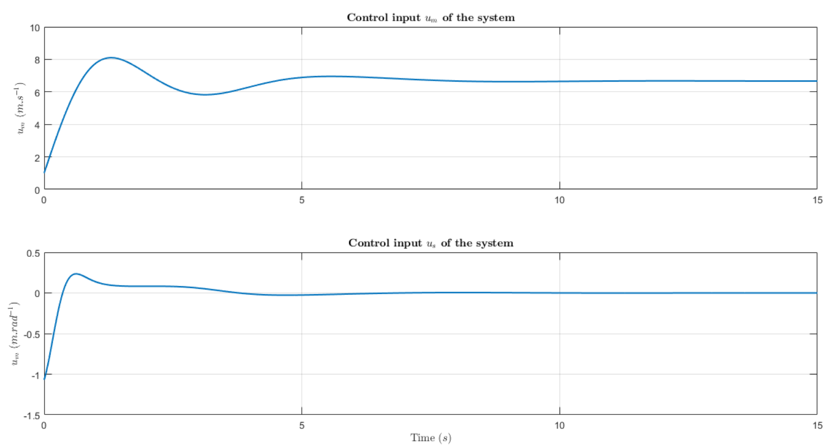

Figure 2.6: The control inputs with $K_x = 1$, $K_y = 1$ and $K_\theta = 1$

- Tuning with $K_x = 10$, $K_y = 10$ and $K_\theta = 10$:

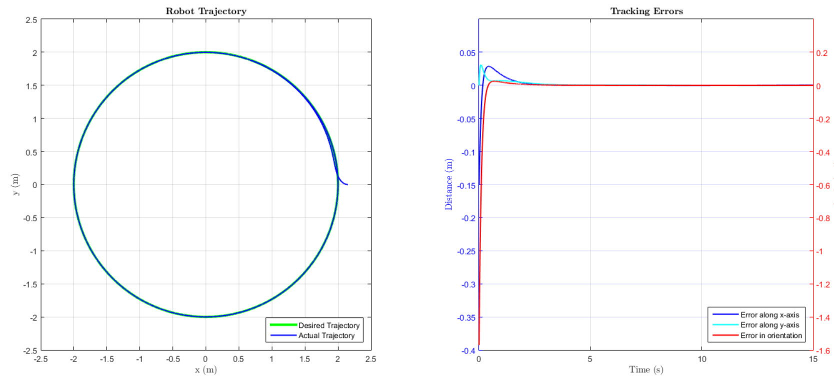

Figure 2.7: Robot Trajectory and errors in R_0 frame with $K_x = 10$, $K_y = 10$ and $K_\theta = 10$

As can be noted in the Figure 5.7, the dynamic performance has improved significantly and the error in θ converges to a constant value in the order of 10^{-4} radians but the error in x and y oscillate slowly about zero with a significantly low magnitude in the order of 10^{-5} metres. The settling time is approximately 3.5 seconds. Increasing the gains further may give better results in the simulation but practically the gain cannot be increased indefinitely as high gains can make the system unstable and saturate the actuators.

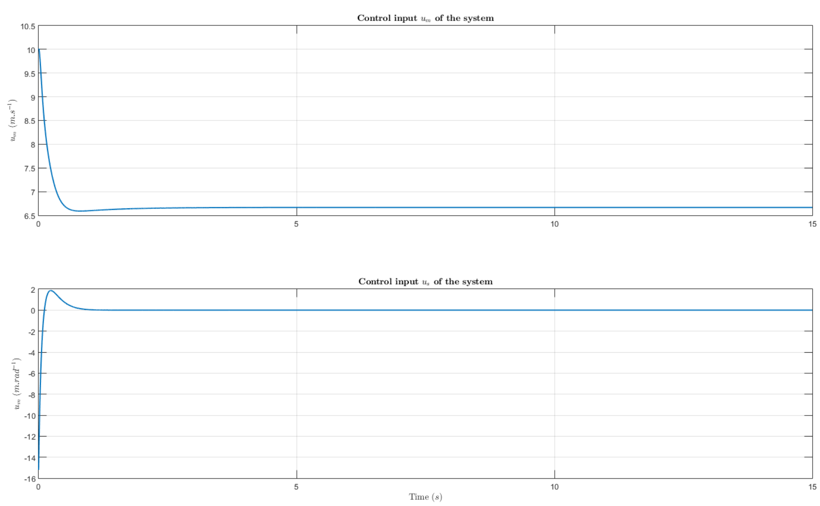

Figure 2.8: Control inputs with $K_x = 10$, $K_y = 10$ and $K_\theta = 10$

Effect of Parameter Variation:

- **Case 1:** 10% increase in value of **a** only

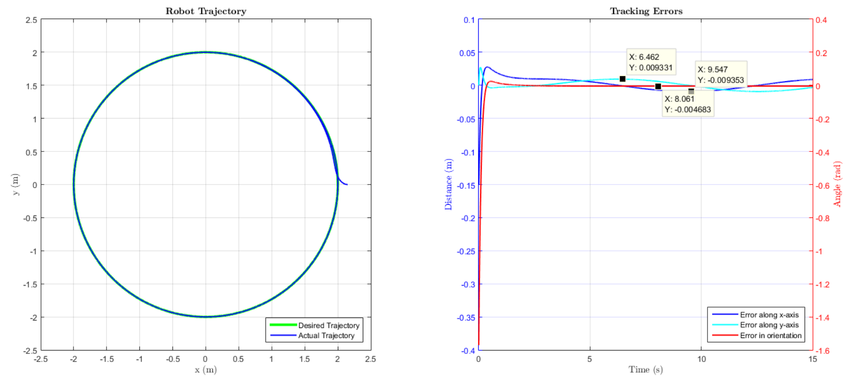

Figure 2.9: Robot Trajectory and errors in R_0 frame with $K_x = 10$, $K_y = 10$ and $K_\theta = 10$ and 10% increase in **a**

The effect of parameter variation can be observed by comparing Figure 5.7 and Figure 5.9. In this case, error in x and y oscillate with a greater amplitude (0.0094 metres and 0.0093 metres respectively) and the error in orientation θ increases to a new steady state value (-0.004683 radians). The overall performance has deteriorated due to parameter variation.

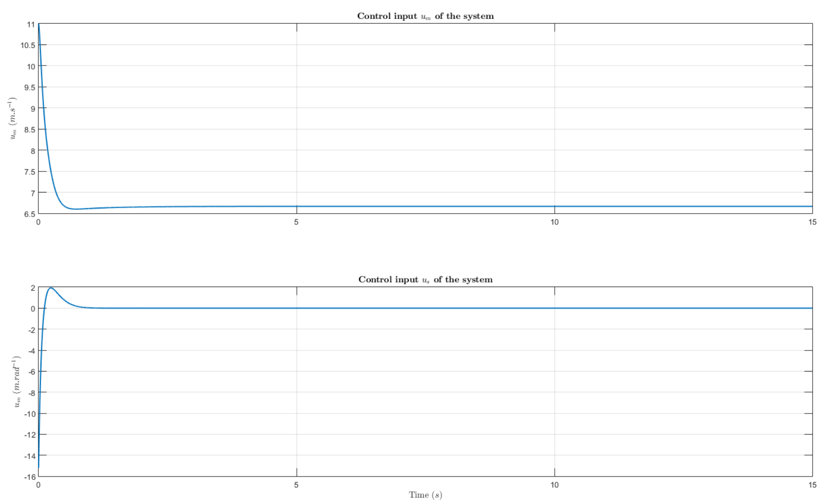

Figure 2.10: Control inputs with $K_x = 10$, $K_y = 10$ and $K_\theta = 10$ and 10% increase in **a**

A possible solution to reduce the effect of parameter variation could be increasing the gains within reasonable limits. On making $K_x = 25$, $K_y = 25$ and $K_\theta = 15$, a reduction in error magnitudes can be seen (Figure 5.11). The errors in x and x now oscillate slowely with a lower amplitude (0.0037 metres in both axes) and the error in orientation θ has also reduced to a lower steadt state value (-0.001881 radians).

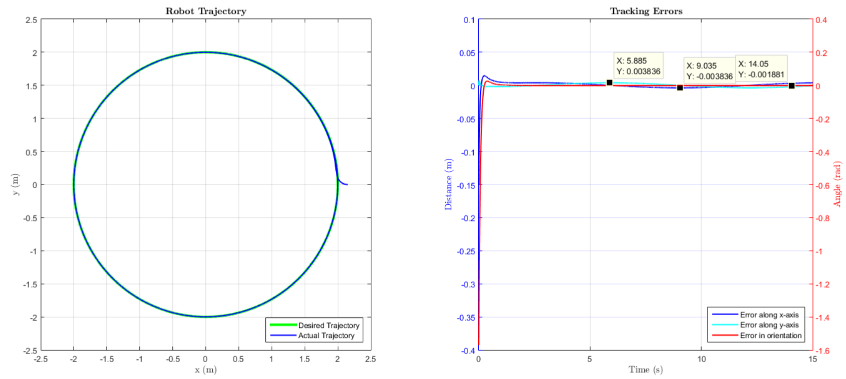

Figure 2.11: Robot Trajectory and errors in R_0 frame with $K_x = 25$, $K_y = 25$ and $K_\theta = 15$ and 10% increase in **a**

- **Case 2:** 10% increase in value of **r** only
  As in Case 1, the effect of variation in **r** results in oscillatory errors in x and y with higher amplitude(approximately 0.009717 metres in both axes) and increased steady state error in θ (approximately 0.004859 radians) as shown in Figure 5.12. A higher gain tends to mitigate this problem as observed in Figure 5.14. The errors in x and y reduce to a lower amplitude (approximately 0.003978 metres in both axes) and converge to a reduced steady state error in θ (approximately 0.001952 radians).

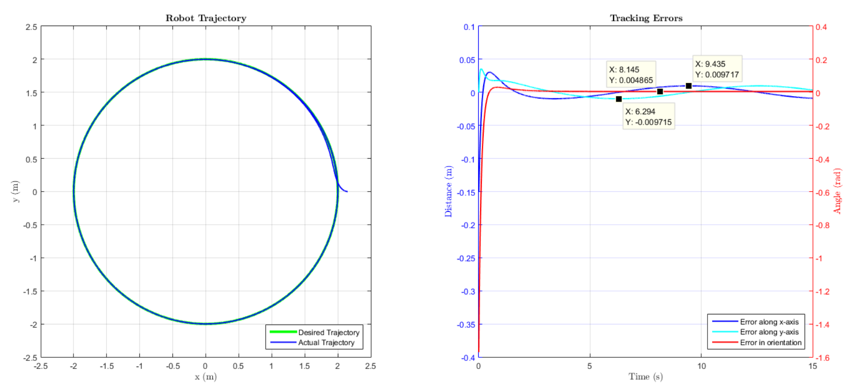

Figure 2.12: Robot Trajectory and errors in R_0 frame with $K_x = 10$, $K_y = 10$ and $K_\theta = 10$ and 10% increase in **r**

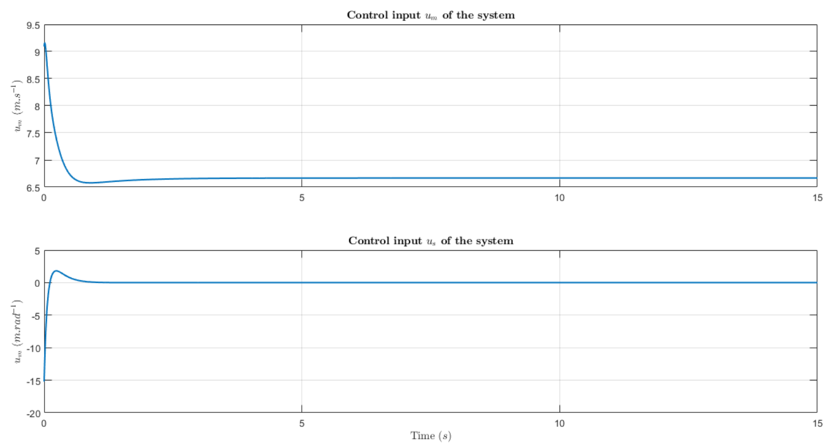

Figure 2.13: Control inputs with $K_x = 10$, $K_y = 10$ and $K_\theta = 10$ and 10% increase in $\mathbf{r}$

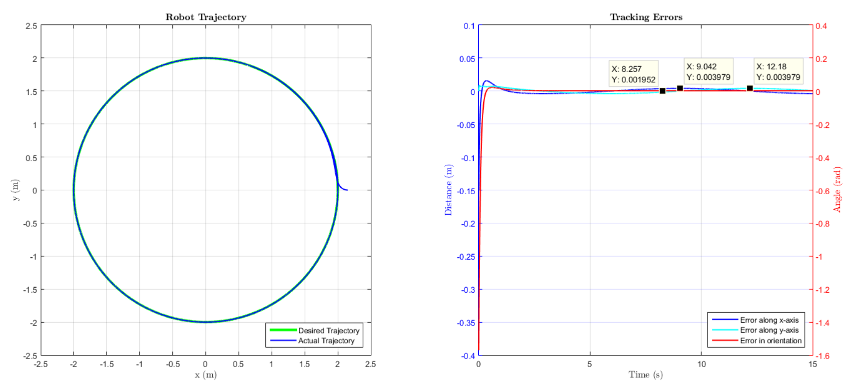

Figure 2.14: Robot Trajectory and errors in R_0 frame with $K_x = 25$, $K_y = 25$ and $K_\theta = 15$ and 10% increase in $\mathbf{r}$

The effect of 10% decrease in the above parameters gives similar results but with changes in magnitude and phases, nevertheless, the errors reduced when a higher gains are applied.