A dual Ackerman drive would steer both front and rear wheels using an Ackerman steering approach. What would the pros and cons for this system compared to a single Ackerman drive?

Pros

- -can still be used easily and intuitively with a conventional steering wheel
- -gives steering over all 4 wheels
- -allows for tighter turns about some pivot point not centered on the vehicle
- -drive simillar to most newer lawn tractors

Cons:

-increased complexity means higher liklihood of failure

Real motion and measurement involves error and this problem will introduce the concepts. Assume that you have a differential drive robot with wheels that are 20cm in radius and L is 12cm. Using the differential drive code (forward kinematics) from the text, develop code to simulate the robot motion when the wheel velocities are ϕ 1=0.25t2, ϕ 2=0.5t. The starting location is [0,0] with θ =0

- 1. Plot the path of the robot on $0 \le t \le 5$. It should end up somewhere near [50,60].
- 2. Assume that you have Gaussian noise added to the omegas each time you evaluate the velocity (each time step). Test with μ =0 and σ =0.3. Write the final location (x,y) to a file and repeat for 100 simulations. Hint:

```
mu, sigma = 0.0, 0.3

xerr = np.random.normal(mu,sigma, NumP)

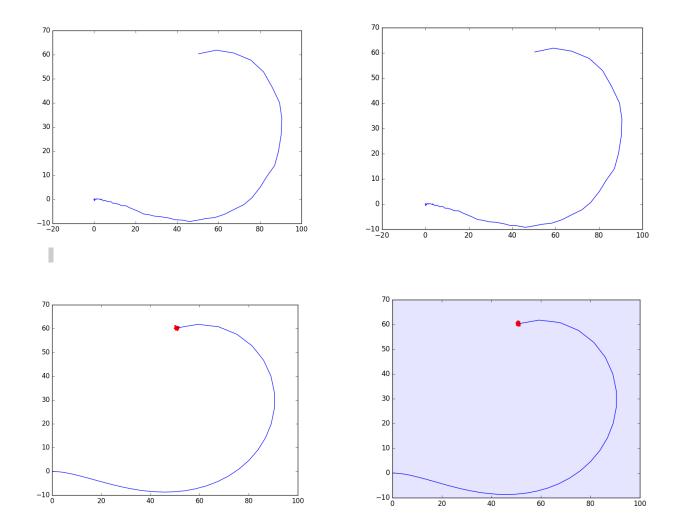
yerr = np.random.normal(mu,sigma, NumP)
```

3.Generate a plot that includes the noise free robot path and the final locations for the simulations with noise. Hint:

```
import numpy as np
import pylab as plt
...
plt.plot(xpath,ypath, 'b-', x,y, 'r.')
plt.xlim(-10, 90)
plt.ylim(-20, 80)
plt.show()
```

4.Find the location means and 2x2 covariance matrix for this data set, and compute the eigenvalues and eigenvectors of the matrix. Find the ellipse that these generate. [The major and minor axes directions are given by the eigenvectors. Show the point cloud of final locations and the ellipse in a graphic (plot the data and the ellipse). Hint:

```
from scipy import linalg
from matplotlib.patches import Ellipse
# assume final locations are in x \& y
mat = np.array([x,y])
# find covariance matrix
cmat = np.cov(mat)
# compute eigenvals and eigenvects of covariance
eval, evec = linalg.eigh(cmat)
# find ellipse rotation angle
angle = 180*atan2(evec[0,1],evec[0,0])/np.pi
# create ellipse
ell = Ellipse((np.mean(x),np.mean(y)),
                 eval[0],eval[1],angle)
# make the ellipse subplot
a = plt.subplot(111, aspect='equal')
ell.set_alpha(0.1) # make the ellipse lighter
a.add_artist(ell) # add this to the plot
```



Top Left: 6.5.a Top Right: 6.5.b Bottom Left: 6.5.c Bottom Right: 6.5.d

```
Code Base for problem 6.5:
a.)
        import pylab as plt
import numpy as np
from math import *
N = 52
x = np.zeros(N)
y = np.zeros(N)
q = np.zeros(N)
x[0] = 0; y[0] = 0; q[0] = 0.0
t = 0; dt = 0.1
def ddstep(xc, yc, qc,r,l,dt,w1,w2):
  xn = xc + (r*dt/2.0)*(w1+w2)*cos(qc)
 yn = yc + (r*dt/2.0)*(w1+w2)*sin(qc)
 qn = qc + (r*dt/(2.0*l))*(w1-w2)
 return (xn,yn,qn)
for i in range(N-1):
 w1 = 0.25*t*t
  w2 = 0.5*t
 x[i+1], y[i+1], q[i+1] = ddstep(x[i], y[i], q[i], 20,12.0,dt,w1,w2)
 t = t + dt
 print(t)
plt.plot(x,y,'b')
plt.show()
b.)
import pylab as plt
import numpy as np
from math import *
N = 52
mu, sigma = 0.0, 0.3
filename = "output.txt"
file = open(filename, "w")
for k in range(100):
  x = np.zeros(N)
  y = np.zeros(N)
  x_with_error = np.zeros(N)
  y with error = np.zeros(N)
  q = np.zeros(N)
  xerr = np.random.normal(mu,sigma, 100)
  yerr = np.random.normal(mu,sigma, 100)
  x[0] = 0; y[0] = 0; q[0] = 0.0
  t = 0; dt = 0.1
  def ddstep(xc, yc, qc,r,l,dt,w1,w2):
     xn = xc + (r*dt/2.0)*(w1+w2)*cos(qc)
```

yn = yc + (r*dt/2.0)*(w1+w2)*sin(qc)qn = qc + (r*dt/(2.0*l))*(w1-w2)

```
return (xn,yn,qn)
  for i in range(N-1):
     w1 = 0.25*t*t
     w2 = 0.5*t
     x_with_error[i] = xerr[i] + x[i]
     y_with_error[i] = yerr[i] + y[i]
     x[i+1], y[i+1], q[i+1] = ddstep(x[i], y[i], q[i], 20,12.0, dt, w1, w2)
     x\_with\_error[i+1], y\_with\_error[i+1], q[i+1] = ddstep(x\_with\_error[i], y\_with\_error[i], q[i], 20, 12.0, dt, w1, w2)
     t = t + dt
     print(t)
  file.write('\{0\} \{1\}\n'.format(x[51], y[51]))
  k+=1
file.close()
plt.plot(x_with_error,y_with_error,'b')
plt.show()
c.)
import pylab as plt
import numpy as np
from math import *
N = 52
mu, sigma = 0.0, 0.3
filename = "output.txt"
file = open(filename, "w")
for k in range (100):
  x = np.zeros(N)
  y = np.zeros(N)
  x_with_error = np.zeros(N)
  y_with_error = np.zeros(N)
  q = np.zeros(N)
  xerr = np.random.normal(mu,sigma, 100)
  yerr = np.random.normal(mu,sigma, 100)
  x[0] = 0; y[0] = 0; q[0] = 0.0
  t = 0; dt = 0.1
  def ddstep(xc, yc, qc,r,l,dt,w1,w2):
     xn = xc + (r*dt/2.0)*(w1+w2)*cos(qc)
     yn = yc + (r*dt/2.0)*(w1+w2)*sin(qc)
     qn = qc + (r*dt/(2.0*1))*(w1-w2)
     return (xn,yn,qn)
  for i in range(N-1):
     w1 = 0.25*t*t
     w2 = 0.5*t
     x_with_error[i] = xerr[i] + x[i]
```

```
y_with_error[i] = yerr[i] + y[i]
     x[i+1], y[i+1], q[i+1] = ddstep(x[i], y[i], q[i], 20,12.0,dt,w1,w2)
     x_{\text{with\_error}[i+1]}, y_{\text{with\_error}[i+1]}, q[i+1] = ddstep(x_{\text{with\_error}[i]}, y_{\text{with\_error}[i]}, q[i], 20, 12.0, dt, w1, w2)
     t = t + dt
  plt.plot(x_with_error[51],y_with_error[51],'r.')
  k+=1
plt.plot(x,y,'b-')
plt.show()
d.)
import pylab as plt
import numpy as np
from scipy import linalg
from matplotlib.patches import Ellipse
from math import *
N=52
mu, sigma = 0.0, 0.3
filename = "output.txt"
file = open(filename, "w")
x final = np.zeros(100)
y_final = np.zeros(100)
for k in range (100):
        x = np.zeros(N)
        y = np.zeros(N)
        x_with_error = np.zeros(N)
        y_with_error = np.zeros(N)
        q = np.zeros(N)
        xerr = np.random.normal(mu,sigma, 100)
        yerr = np.random.normal(mu,sigma, 100)
        x mean = 0
        y_mean = 0
        xerr_mean = 0
        yerr_mean = 0
        x[0] = 0; y[0] = 0; q[0] = 0.0
        t = 0; dt = 0.1
        def ddstep(xc, yc, qc,r,l,dt,w1,w2):
                 xn = xc + (r*dt/2.0)*(w1+w2)*cos(qc)
                 yn = yc + (r*dt/2.0)*(w1+w2)*sin(qc)
                 qn = qc + (r*dt/(2.0*l))*(w1-w2)
                 return (xn,yn,qn)
        for i in range(N-1):
                 w1 = 0.25*t*t
                 w2 = 0.5*t
                 x_with_error[i] = xerr[i] + x[i]
```

```
y_with_error[i] = yerr[i] + y[i]
                x_mean += x[i]
                y_mean += y[i]
                xerr_mean += x_with_error[i]
                yerr_mean += y_with_error[i]
                x_with_error[i+1], y_with_error[i+1], q[i+1] = ddstep(x_with_error[i], y_with_error[i],
q[i],20,12.0,dt,w1,w2)
                x[i+1], y[i+1], q[i+1] = ddstep(x[i], y[i], q[i], 20,12.0,dt,w1,w2)
                t = t + dt
        x_{final}[k] = x_{with\_error}[51]
        y_{\text{final}[k]} = y_{\text{with\_error}[51]}
x mean = N
y_mean /= N
xerr_mean /= N
yerr_mean /= N
print(x_mean)
print(y_mean)
tempx = (x_mean + xerr_mean)/2
tempy = (y_mean + yerr_mean)/2
varx = (x_mean - tempx)*(x_mean - tempx) + (xerr_mean - tempx)*(xerr_mean - tempx)
vary= (y_mean - tempy)*(y_mean - tempy) + (yerr_mean - tempy)*(yerr_mean - tempy)
covxy= (x_mean - tempx)*(y_mean - tempy) + (xerr_mean - tempx)*(yerr_mean - tempy)
# assume final locations are in x & y
mat = np.array([x,y])
# find covariance matrix
cmat = np.cov(mat)
# compute eigenvals and eigenvects of covariance
eval, evec = linalg.eigh(cmat)
# find ellipse rotation angle
angle = 180*atan2(evec[0,1],evec[0,0])/np.pi
# create ellipse
ell = Ellipse((np.mean(x),np.mean(y)),
                         eval[0],eval[1],angle)
# make the ellipse subplot
a = plt.subplot(111, aspect='equal')
ell.set alpha(0.1) # make the ellipse lighter
a.add_artist(ell) # add this to the plot
print(eval)
for j in range(100):
        plt.plot(x_final[j], y_final[j],'r.')
plt.plot(x,y,'b-')
plt.show()
```