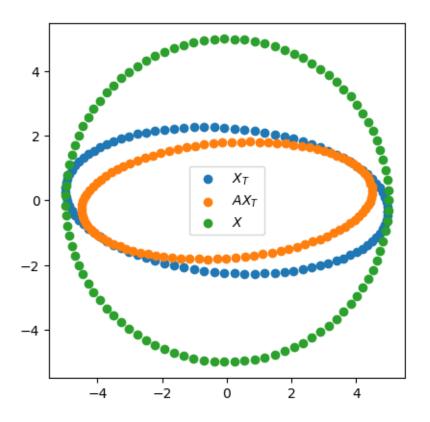
## semi\_definite\_program

## May 19, 2024

```
[1]: import cvxpy as cvx
     import matplotlib.pyplot as plt
     import numpy as np
     def generate_ellipsoid_points(M, num_points=100):
         """Generate points on a 2-D ellipsoid.
         The ellipsoid is described by the equation
         `{ x \mid x.T \otimes inv(M) \otimes x \le 1 }`,
         where inv(M) denotes the inverse of the matrix argument M.
         The returned array has shape (num_points, 2).
         L = np.linalg.cholesky(M)
          = np.linspace(0, 2*np.pi, num_points)
         u = np.column_stack([np.cos(), np.sin()])
         x = u @ L.T
         return x
[2]: A = np.array([[0.9, 0.6], [0, 0.8]])
     B = np.array([[0],[1]])
     r_x = 5
     ru = 1
     Q = np.eye(2)
     R = np.eye(1)
     P = np.eye(2)
     N = 4
     A, B
[2]: (array([[0.9, 0.6],
             [0., 0.8]]),
      array([[0],
             [1]]))
```

```
[3]: M = cvx.Variable((2,2),symmetric=True)
     constraints = []
     constraints += [M<<r_x**2 * np.eye(2)]
     constraints += [cvx.bmat([[M, A@M],[M@A.T, M]])>>0]
     prob = cvx.Problem(
         cvx.Maximize(cvx.log_det(M)),
         constraints
[4]: prob.solve()
[4]: 4.832662985874696
[5]: M. value
[5]: array([[24.86408695, -1.64219805],
            [-1.64219805, 5.15770616]])
[6]: np.linalg.inv(M.value)
[6]: array([[0.04108258, 0.01308057],
            [0.01308057, 0.19804945]])
[7]: ellipsoid_XT = generate_ellipsoid_points(M.value)
     ellipsoid_A_XT = np.array([A@x for x in ellipsoid_XT])
      = np.linspace(0, 2*np.pi, 100)
     ellipsoid_X = r_x*np.column_stack([np.cos(), np.sin()])
[8]: fig, ax = plt.subplots()
     ax.scatter(ellipsoid_XT[:,0],ellipsoid_XT[:,1], label='$X_T$')
     ax.scatter(ellipsoid_A_XT[:,0],ellipsoid_A_XT[:,1], label='$AX_T$')
     # ax.scatter(np.array([A@x for x in ellipsoid_A_XT])[:,0],np.array([A@x for x])
     →in ellipsoid_A_XT])[:,1], label='$AAX_T$')
     ax.scatter(ellipsoid_X[:,0],ellipsoid_X[:,1], label='$X$')
     ax.legend()
     ax.set_aspect('equal')
```



## 0.1 The MPC Problem

```
[9]: x0 = np.array([[0],[-4.5]])
x0_cvx = cvx.Parameter((2,1), value=x0)

n, m = Q.shape[0], R.shape[0]

W = np.linalg.inv(M.value)

# Form the CVX Problem
cost = 0.0
constraints = []

x_cvx = cvx.Variable((N + 1, n))
u_cvx = cvx.Variable((N, m))

# Define Cost
cost += cvx.QuadForm(x_cvx[-1,:], P)
for i in range(N):
    cost += cvx.QuadForm(x_cvx[i,:], Q)
    cost += cvx.QuadForm(u_cvx[i,:], R)
```

/home/qdeng/.pyenv/versions/3.12.1/envs/AA203/lib/python3.12/sitepackages/cvxpy/reductions/solvers/solving\_chain.py:336: FutureWarning:
 Your problem is being solved with the ECOS solver by default. Starting in
 CVXPY 1.5.0, Clarabel will be used as the default solver instead. To
continue
 using ECOS, specify the ECOS solver explicitly using the ``solver=cp.ECOS``
 argument to the ``problem.solve`` method.

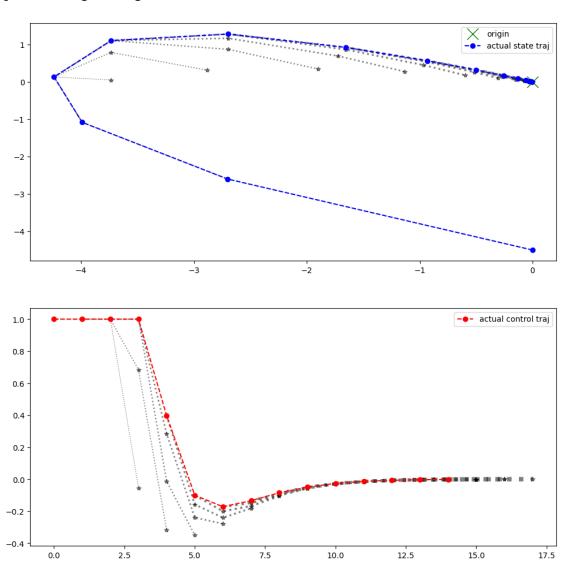
warnings.warn(ECOS\_DEPRECATION\_MSG, FutureWarning)

## [9]: 86.31905795612

```
[11]: T = 15
      x0_cvx.value = x0
      x_mpc = np.zeros((T, N + 1, n))
      u_mpc = np.zeros((T, N, m))
      fig, ax = plt.subplots(2,1, figsize=(12, 12))
      for t in range(T):
          prob.solve()
          x_mpc[t] = x_cvx.value
          u_mpc[t] = u_cvx.value
          # set the parameter value to where the state is at the next time instance
          x0_cvx.value = A @ x0_cvx.value + B * u_mpc[t, 0, :]
          \# Plot out the optimized trajectory for the horizon with length N
          ax[0].plot(x_mpc[t,:,0], x_mpc[t,:,1], 'k*:', linewidth=t*1/3+1, alpha=1/2)
          ax[1].plot(t+np.arange(0,N), u_mpc[t,:,0], 'k*:', linewidth=t*1/3+1,__
       \Rightarrowalpha=1/2)
      ax[0].plot(0,0,'gx',label='origin',markersize=15)
      ax[0].plot(x_mpc[:,0,0], x_mpc[:,0,1], 'bo--', label='actual state traj')
```

```
ax[0].legend()
ax[1].plot(u_mpc[:,0,0], 'ro--', label='actual control traj')
ax[1].legend()
```

[11]: <matplotlib.legend.Legend at 0x7d8a5da37500>



[]: