HW15

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1 HW15

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1.1.1 5.12

```
[1]: import cvxpy as cp
     import numpy as np
     import matplotlib.pyplot as plt
     from ls_perm_meas_data import *
     y=np.reshape(y,m)
     x = cp.Variable(n)
     first_objective = cp.Minimize(cp.norm1(A @ x - y))
     first_problem = cp.Problem(first_objective,[])
     print("first estimate error: ",first_problem.solve())
     rsq = (A@x.value-y)**2
     plt.hist(rsq**0.5);
     sorted_idx = np.argsort(-rsq)
     k_largest_idx = sorted_idx[:k]
     keeping_idx = sorted_idx[k:]
     second_objective = cp.Minimize(cp.norm2(A[keeping_idx,:] @ x - y[keeping_idx]))
     second_problem = cp.Problem (second_objective,[])
     second_problem.solve()
     y_estimate = A[k_largest_idx,:]@x.value
     order_of_y_estimete = np.argsort(y_estimate)
     order_of_y_k_largest = np.argsort(y[k_largest_idx])
     newA = np.zeros(A.shape)
     newA[k:,:]=A[keeping_idx,:]
     newA[:k,:]=A[k_largest_idx,:][order_of_y_estimete]
     newy = np.zeros(y.shape)
     newy[k:]=y[keeping_idx]
     newy[:k]=y[k_largest_idx][order_of_y_k_largest]
     # newy[order_of_y_k_largest]=y[order_of_y_estimete]
```

```
# A[order_of_y_k_largest,:]=A[order_of_y_estimete,:]

final_objective = cp.Minimize(cp.norm1(newA @ x - newy))
final_problem = cp.Problem(final_objective,[])
print("final estimate error: ",final_problem.solve())
rsq = (newA@x.value-newy)**2
plt.figure();
plt.hist(rsq**0.5);
```

first estimate error: 2041.6648818003052 final estimate error: 75.81172584135858

1.1.2 13.3

```
import cvxpy as cp
import numpy as np

G = np.array([[0.3, -0.1, -0.9], [-0.6, 0.3, -0.3], [-0.3, 0.6, 0.2]])
v = np.random.normal(0, 1, 3)
I = np.eye(3)
Z = cp.Variable((3,3))
constrints = [cp.norm(Z,axis=1)<=1]
objective = cp.Minimize(cp.max(cp.norm(I+G@Z,axis=1)))
prob = cp.Problem(objective, constrints)
prob.solve()
print(np.linalg.inv(I+Z.value@G)@Z.value)
print("optimal value is:",np.round(prob.value**2,4))</pre>
```

```
[[-1.30309479 3.68758732 3.45616496]

[ 0.21119619 -0.89652267 -3.36088163]

[15.52709908 7.32358762 -9.1525585 ]]

optimal value is: 0.1647
```

1.1.3 13.4

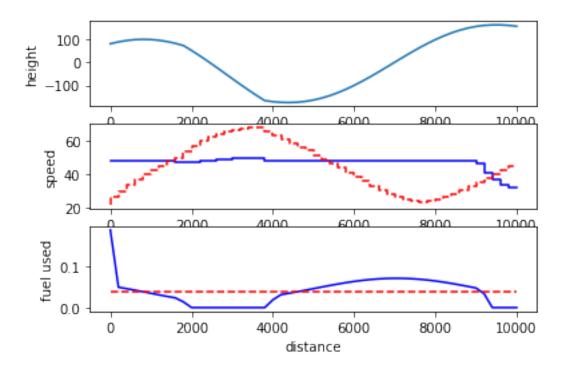
```
[3]: from min_time_speed_data import *
  import matplotlib.pyplot as plt
  import cvxpy as cp
  import numpy as np

f = cp.Variable(N+1)
  f_ob = cp.Variable()
  z = cp.Variable(N+1)
  T = d*sum(cp.inv_pos(cp.sqrt(z[:-1])))

constraints = [f >= 0]
```

```
constraints += [sum(f) + (T*P)/eta <= F]
constraints += [eta*f[0] == 0.5*m*z[0]]
for i in range(N):
    constraints += [0.5*m*z[i+1] + m*g*h[i+1] == 0.5*m*z[i] + m*g*h[i] +_{\sqcup}
→eta*f[i+1] - d*C_D*z[i]]
objective = cp.Minimize(T)
prob = cp.Problem(objective, constraints)
print(prob.solve())
s_opt = np.sqrt(z.value)
f_opt = f.value
prob2 = cp.Problem(objective, constraints + [cp.max(f)*(N+1) <= sum(f)])
print(prob2.solve())
sc = np.sqrt(z.value)
fc = f.value
plt.subplot(3,1,1)
plt.plot(np.array(range(0,N+1))*d,h)
plt.ylabel('height')
plt.subplot(3,1,2)
plt.step(np.array(range(0,N+1))*d,s_opt,'b')
plt.step(np.array(range(0,N+1))*d,sc,'--r')
plt.ylabel('speed')
plt.subplot(3,1,3)
plt.plot(np.array(range(0,N+1))*d, f_opt,'b');
plt.plot(np.array(range(0,N+1))*d, fc,'--r')
plt.xlabel('distance')
plt.ylabel('fuel used')
plt.show()
```

213.2619904455464 258.47941979011



1.1.4 14.8

```
[4]: import cvxpy as cp
     import numpy as np
     p = np.array([1/3, 1/6, 1/3, 1/6])
     R = np.array([[2, 2, 0.5, 0.5], [1.3, 0.5, 1.3, 0.5], [1, 1, 1, 1]])
     x = cp.Variable(3)
     y = cp.Variable(4)
     constraints1 = [y == R.T @ x, sum(x) == 1]
     objective = cp.Minimize(-p @ cp.log(y))
     prob1 = cp.Problem(objective, constraints1)
     prob1.solve()
     print("best strategy is:",x.value, "and optimal value is:",-prob1.value)
     constraints2 = constraints1 + [x[0] == 1, x[1] == 0, x[2] == 0]
     prob2 = cp.Problem(objective, constraints2)
     print("for strategy (1,0,0) : ", -prob2.solve())
     constraints3 = constraints1 + [x[0] == 0, x[1] == 1, x[2] == 0]
     prob3 = cp.Problem(objective, constraints3)
     print("for strategy (0,1,0) : ", -prob3.solve())
     constraints4 = constraints1 + [x[0] == 0.5, x[1] == 0.5, x[2] == 0]
     prob4 = cp.Problem(objective, constraints4)
     print("for strategy (0.5,0.5,0) : ", -prob4.solve())
     constraints5 = constraints1 + [x[0] == 0, x[1] == 0, x[2] == 1]
```

```
prob5 = cp.Problem(objective, constraints5)
print("for strategy (0,0,1) : ", -prob5.solve())

best strategy is: [0.49725105 0.19936803 0.30338093] and optimal value is:
0.06227157878135451
for strategy (1,0,0) : -4.923437657700523e-10
for strategy (0,1,0) : -0.056139550858751545
for strategy (0.5,0.5,0) : 0.05347098525410708
for strategy (0,0,1) : -1.8126598794172385e-10
```

1.1.5 14.10

```
[5]: import cvxpy as cp
     import numpy as np
     from opt_funding_data import *
     x = cp.Variable(n)
     B0 = cp.Variable()
     B = cp.Variable(T)
     I = A @ x
     Sp = cp.Variable(T-1)
     Sn = cp.Variable(T-1)
     objective = cp.Minimize( P @ x + B0 )
     constraints = [x \ge 0, B0 \ge 0, B[0] = (1+rp)*B0, B[T-1]+I[T-1]-E[T-1]=0]
     constraints += [Sp>=0,Sn>=0,B[1:]==(1+rp)*Sp -(1+rn)*Sn , B[:-1]+I[:-1]-E[:
     \rightarrow-1]==Sp-Sn]
     problem1=cp.Problem(objective,constraints)
     problem1.solve()
     print("optimal investment is:",np.round(problem1.value,4))
     print("optimal x is:",np.round(x.value,4))
     plt.plot(np.append(B0.value,B.value),label="with bond");
     problem2 = cp.Problem(objective,constraints+[x==0])
     problem2.solve()
     print("optimal investment without bond is:",np.round(problem2.solve(),4))
     plt.plot(np.append(B0.value,B.value),label="no bond");
    plt.legend();
```

```
optimal investment is: 40.7495 optimal x is: [ 0. 18.9326 0. 0. 13.8493 8.9228] optimal investment without bond is: 41.7902
```

