

Columbia University  
COMSW4771  
Dr. James McInerney  
Homework 0

Submitted By: Alexander Stein, as5281

September 10th, 2017

Problem 1.1

(i) What is the marginal distribution of X?

Table 1: Marginal Probabilities

	Y1	Y2	Y3	P <sub>x</sub> (i)
X1	0.1	0.2	0.3	0.6
X2	0.2	0.1	0.1	0.4
P <sub>y</sub> (j)	0.3	0.3	0.4	1.00

(ii) What is  $\Pr[Y=1 \mid X=2]$ ?

$$P(Y=1|X=2) = P(Y=1, X=2) / P(X=2)$$

$$P(Y=1|X=2) = (0.2)/(0.4)$$

$$P(Y=1|X=2) = 1/2$$

(iii) Let  $f : x \rightarrow x^2$ . What is  $E[f(x)|Y=1]$ ?

$$E(x|Y = y)$$

$$= \sum_{x \in X} xP(X = x|Y = y)$$

$$= \sum_{x \in X} xP(x = x, Y = y)/P(Y = y)$$

$$\begin{aligned} & E(f(x)|Y = 1) \\ &= \sum_{x \in X} x^2 P(X = x, Y = 1)/P(Y = 1) \\ &= (1^2) * (0.1)/(0.3) + (2^2) * (0.2)/(0.3) \\ &= 1/3 + 8/3 \\ &= 3 \end{aligned}$$

### Problem 1.2

(i) Verify that  $\frac{1}{\theta}e^{-x/\theta}$  is a probability distribution.

To verify, we will show that the integral from 0 to  $\infty$  is 1

$$\begin{aligned} & \int_0^\infty \frac{1}{\theta} e^{x/\theta} dx \\ & u = -\frac{x}{\theta} \rightarrow dx = -\theta du \\ &= -\frac{1}{\theta} \int_0^\infty \theta e^u du \\ &= -e^u = -e^{-x/\theta} \Big|_0^\infty \\ &= (-e^{-\infty/\theta}) - (-e^{-0/\theta}) \\ &= 1 \end{aligned}$$

(ii) What is  $E[X]$  ?

$$\begin{aligned} \mu = E[X] &= \int_{-\infty}^\infty xf(x)dx = \frac{1}{\theta} \int_0^\infty e^{-x/\theta} \\ & u = x/\theta dx = \theta du \\ &= \theta \int ue^{-u} du \\ & f = u \text{ and } g' = e^{-u} \\ & f' = 1 \text{ and } g = -e^{-u} \end{aligned}$$

$$\begin{aligned}
&= fg - \int f'g \rightarrow -u\theta e^{-u} + \theta \int e^{-u} du \\
&= -ue^{-u} - e^{-u} = -xe^{-x/\theta} - \theta e^{-x/\theta} \Big|_0^\infty \\
&= (0) - (-\theta) = \theta
\end{aligned}$$

(iii) What is  $\text{Var}(X)$  ?

$$\text{Var}(X) = E((X - \mu)^2) = E(X^2) - E(X)^2 = E(X^2) - \mu^2$$

so all we need to find is  $E(X^2)$

$$\begin{aligned}
E(X^2) &= \int_0^\infty \frac{x^2}{\theta} e^{-x/\theta} dx \\
u &= x/\theta \text{ and } dx = \theta du \\
&= \theta^2 \int u^2 e^{-u} du \\
f &= u^2 \text{ and } g' = e^{-u} \\
f' &= 2u \text{ and } g = -e^{-u} \\
&= fg - \int f'g \rightarrow \theta^2 [-u^2 e^{-u} - \int_0^\infty -2ue^{-u} du] \\
f_1 &= u \text{ and } g'_1 = e^{-u} \\
f'_1 &= 1 \text{ and } g_1 = -e^{-u} \\
&= \theta^2 [-u^2 e^{-u} + 2[-ue^{-u} + \int_0^\infty -e^{-u} du]] \\
&= \theta^2 [-u^2 e^{-u} - 2ue^{-u} - 2e^{-u}] \\
&= -(x^2 e^{-x/\theta} + 2\theta x e^{-x/\theta} + 2\theta^2 e^{-x/\theta}) \Big|_0^\infty \\
&= (0) - (-2\theta^2) = 2\theta^2
\end{aligned}$$

$$\text{Var}(X) = E(X^2) - E(X)^2 = 2\theta^2 - \theta^2 = \theta^2$$

Problem 1.3

Table 2: Burglary Probabilities

	Robbery(5%)	No Robbery(95%)
Alarm	99%	10%
No Alarm	1%	90%

We are effectively using Bayes Theorem to solve this problem:

$$\begin{aligned}
 P(\text{"true\_alarm"}) &= P(\text{Robbery}) * P(\text{Alarm} \mid \text{Robbery}) \\
 &= (0.05) * (0.99) = 0.0495 \\
 P(\text{"false\_alarm"}) &= P(\text{No\_Robbery}) * P(\text{Alarm} \mid \text{No\_Robbery}) \\
 &= (0.95) * (0.10) = 0.095 \\
 P(\text{"any\_alarm"}) &= P(\text{"true\_alarm"}) + P(\text{"false\_alarm"}) \\
 &= 0.1445
 \end{aligned}$$

$$P(\text{Robbery} \mid \text{"any\_alarm"}) = \frac{P(\text{"true\_alarm"})}{P(\text{"any\_alarm"})} = \frac{0.0495}{0.1445} = 0.34256$$

Problem 2.1

Subspace S is spanned by  $a_1 = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$ ,  $a_2 = \begin{bmatrix} 2 \\ 8 \\ 3 \\ 2 \end{bmatrix}$ ,  $a_3 = \begin{bmatrix} 3 \\ 10 \\ 6 \\ 6 \end{bmatrix}$ .

(i) what is the dimension of the subspace S?

we can see that  $a_1 + a_2 - a_3 = 0$ , But there is no scalar multiple  $k$  such that  $ka_1 = a_2$ . Thus,  $a_1$  and  $a_2$  are linearly independent, and the dimension of  $S$  is 2.

(ii) Project the vector  $\begin{vmatrix} 6 \\ 5 \\ 9 \\ 2 \end{vmatrix}$  onto the subspace  $S$

First we need an orthogonal basis, we will use Gram-Schmidt Process

$$\begin{aligned} v_1 &= a_1 \\ v_2 &= a_2 - \frac{(a_2, v_1)}{\|v_1\|^2} v_1 \\ v_2 &= \begin{vmatrix} 5/6 \\ 17/3 \\ -1/2 \\ -8/3 \end{vmatrix} \end{aligned}$$

Now we can project...

$$proj_s u = \frac{(u, v_1)}{\|v_1\|^2} v_1 + \frac{(u, v_2)}{\|v_2\|^2} v_2 = 1.7v_1 + 0.58v_2$$

Problem 2.2 - SKIPPED, don't know how

Problem 3.1 - SKIPPED, don't know how

Problem 4.1 (see attached images for plots)

CODE:

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.io import loadmat
from scipy import linalg as LA

mat_dict = {}
mat_dict.update(loadmat('hw0data.mat'))

M = mat_dict['M']

print "The dimensions of matrix M: ", M.shape
print "The 4th row of M: ", M[3,:]
print "The 5th column of M: ", M[:,4]
print "Mean value of the 5th column of M: ", np.mean(M[:,4])

print "histogramming the 4th row..."
plt.hist(M[3,:], rwidth=0.5)
plt.title("Histogram of M[3,:]")
plt.xlabel("Value")
plt.ylabel("Frequency")
plt.show()

M_square = np.dot(np.transpose(M), M)
e_vals, e_vecs = LA.eig(M_square)

print "Top three eigenvalues of M_transpose * M: ", e_vals[0:3]
```

Problem 4.2 (see attached images for plots)

CODE:

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.io import loadmat
from scipy import linalg as LA
```

```

L = np.matrix('1.25 -1.5; -1.5 5')

print L

# 500 2-vectors from Gaussian distribution, normalized
R = np.random.normal(0,1,(500,2))
for i, row in enumerate(R):
    R[i,:] = row / np.linalg.norm(row)

# Distorted Vectors
R_hat = np.dot(R,L)

# Eigenvalues of L
e_vals, e_vecs = LA.eig(L)
lambda_max = e_vals.max()
lambda_min = e_vals.min()
v_max = e_vecs[1];
print "Lambda Max: ", lambda_max
print "V_max: ", v_max
print "Lambda Min: ", lambda_min

# Magnitudes of R_hat
R_hat_mag = np.array([np.linalg.norm(row) for row in R_hat])

# Histogram of Distorted Vectors' Magnitudes
print "histogramming Distorted Vectors Magnitudes"
plt.hist(R_hat_mag, rwidth=0.1, bins=np.arange(0.6, 5.7, 0.1))
plt.title("Histogram of Distorted Vectors")
plt.xlabel("Value")
plt.ylabel("Frequency")
plt.show()

# Plot of All the Distorted Vectors and compare to lambda_max
print "plotting Distorted Vectors vs. V_max"
rows,cols = R_hat.shape

for i,l in enumerate(range(0,rows)):

```

```

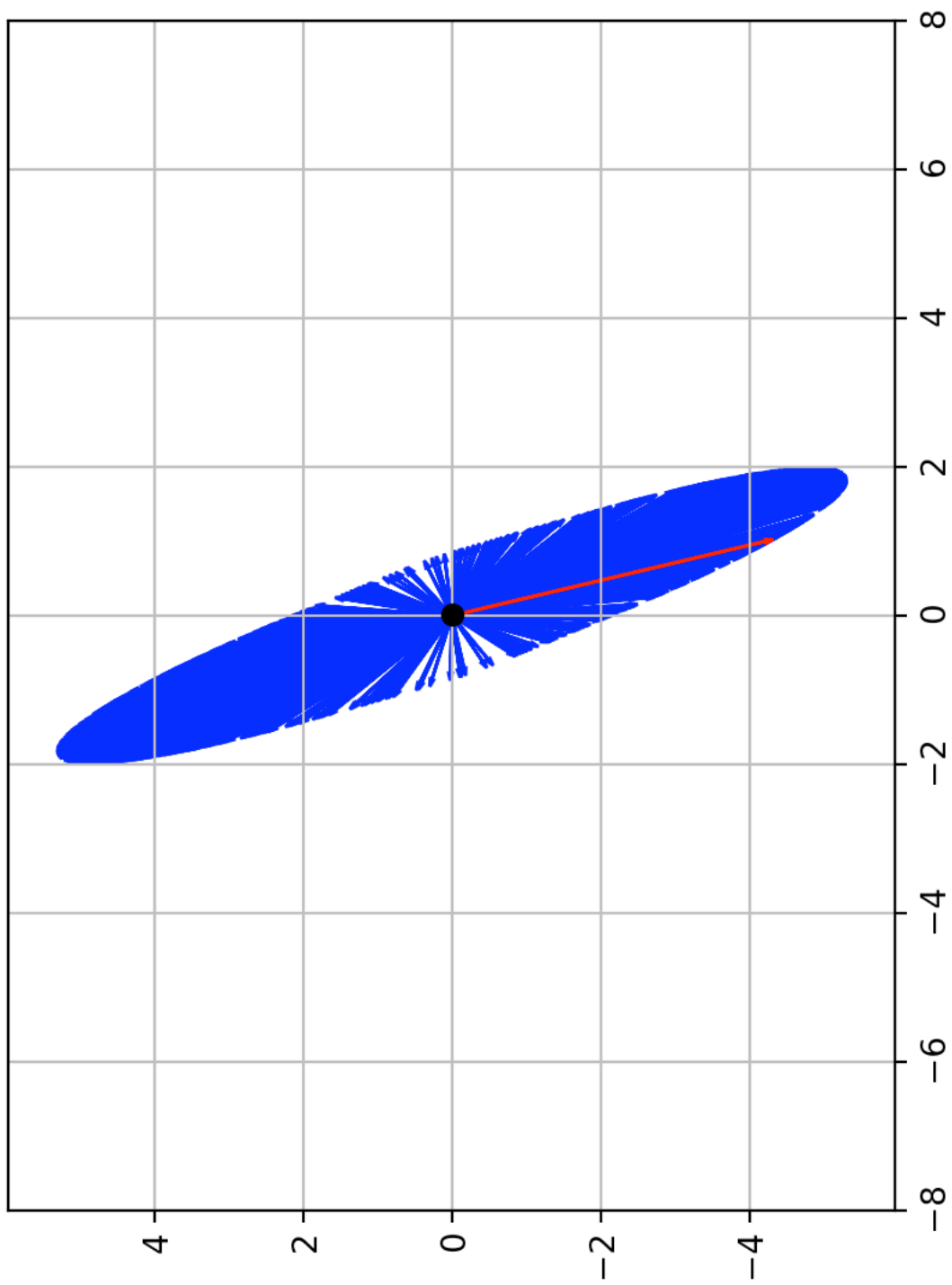
plt.axes().arrow(0,0,R_hat[i,0], \
    R_hat[i,1],head_width=0.05,head_length=0.1,color = 'b')

l_v_max = np.dot(L,v_max)
plt.axes().arrow(0,0,l_v_max[0,0], \
    l_v_max[0,1],head_width=0.05,head_length=0.1,color = 'r')

plt.plot(0,0,'ok') #<-- plot a black point at the origin
plt.axis('equal') #<-- set the axes to the same scale
plt.xlim([-8,8]) #<-- set the x axis limits
plt.ylim([-8,8]) #<-- set the y axis limits
plt.grid(b=True, which='major') #<-- plot grid lines
plt.show()

```





Histogram of Distorted Vectors

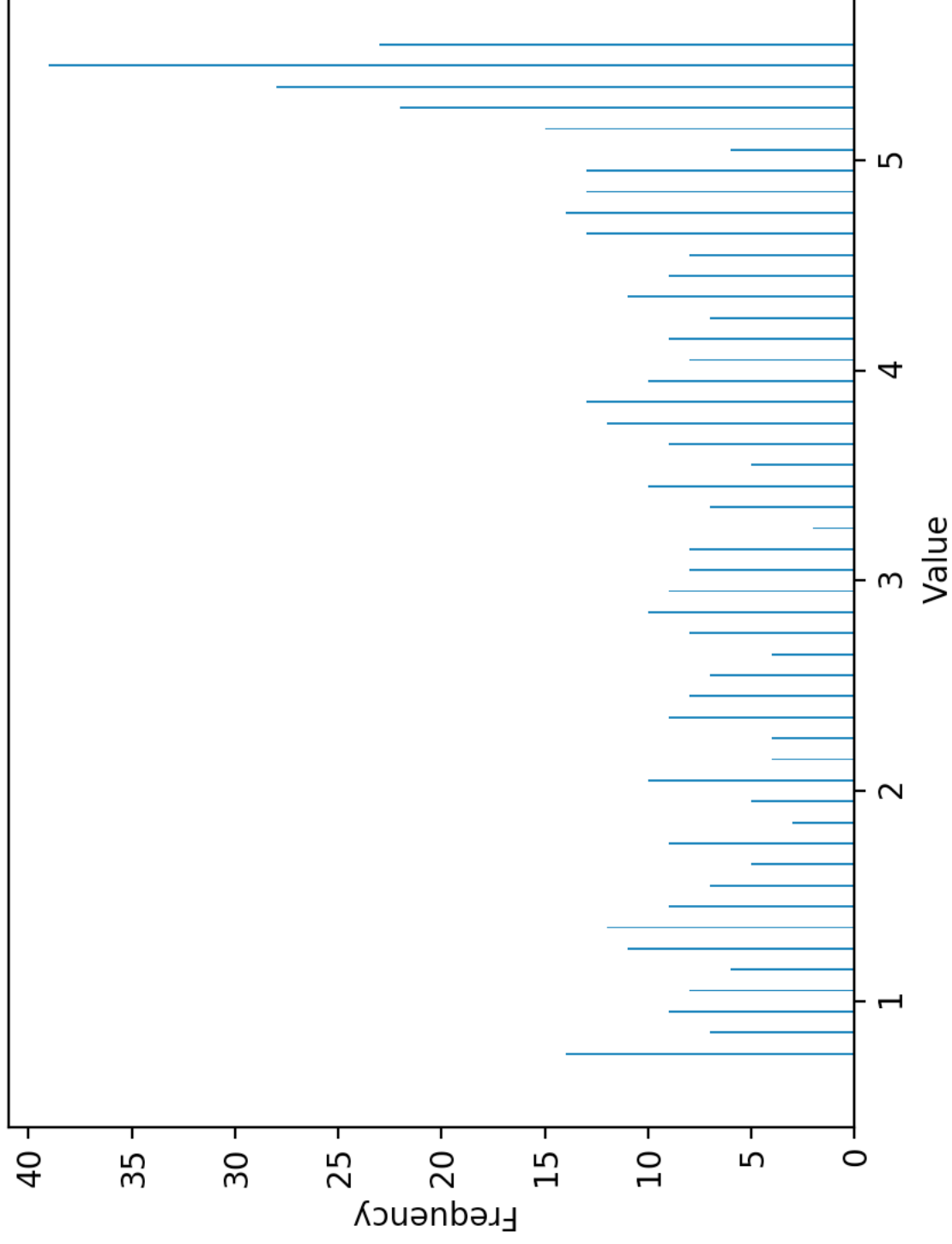


Figure 1

