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Math 18 B01

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Exercise 5.1:

A)

Setting Vectors:

v =

2 0 -1

w =

1 3 3

>> x = [6; 1; -3]

χ =

6 1 -3

>> y = [1; 0; 2]

y =

1 0 2

>> z = [2; -15; -1]

z =

2 -15 -1

Based on our data, the orthogonal subsets include $\{v, y\}$, $\{w, z\}$, $\{x, y\}$, $\{x, z\}$ and $\{y, z\}$. The maximum number of nonzero orthogonal vectors that you can possibly find in R^3 is 3. This is because if vectors are orthogonal to each other, they are also linearly independent. And because our subset is in R^3, we can have at most 3 pivotal columns for an augmented matrix of vectors.

```
b)
>> x = x/norm(x); y = y/norm(y); z = z/norm(z);
>> W = [x y z]

W =

0.8847   0.4472   0.1319
0.1474   0  -0.9891
-0.4423   0.8944  -0.0659
```

Exercise 5.2

Multiplying a matrix by its transpose gets us the identity matrix.

```
b)
>> a = [1;1;0]
a =
  1
   1
  0
>> b = [2;0;3]
b =
  2
  3
>> norm(b)
  3.6056
>> norm(W*b)
ans =
  3.6056
The norms are the same.
>> dot(a, b)
ans =
  2
>> dot(W*a, W*b)
ans =
```

2

```
c)
>> inv(W)
ans =
  0.8847  0.1474  -0.4423
  0.4472 0 0.8944
  0.1319 -0.9891 -0.0659
>> W'
ans =
  0.8847  0.1474  -0.4423
  0.4472 0 0.8944
  0.1319 -0.9891 -0.0659
We can see that the inverse of W and W' are the same matrix.
Exercise 5.3
a)
>> vbar = (dot(v, w)/dot(w, w))*w
vbar =
 -0.0526
 -0.1579
 -0.1579
>> z = v - vbar
z =
  2.0526
  0.1579
 -0.8421
>> vsum = z + vbar
vsum =
  2
```

0 -1

```
b)
>> dot(z, vbar)
ans =
    -2.7756e-17
```

When analyzing this value, we can conclude that z is orthogonal to vbar. This is because although the value isn't exactly zero, there are expected computational rounding errors that may cause a negligible change from zero.

Exercise 5.4

```
>> z = [3;3;3];

>> (dot(z,x)/dot(x,x))*x+(dot(z,y)/dot(y,y))*y

ans =

3.3652

0.2609

2.8174
```

Exercise 5.5

```
>> a = [1; 2; 1]
```

a =

1

2

>> b = [2; 1; 2]

b =

2

1

2

>> c = [1; 1; 2]

```
c =
  1
  1
  2
>> set = [a b c]
set =
  1 2 1
  2
     1 1
      2
          2
>> [Q, R] = qr(set)
Q =
 -0.4082 0.5774 -0.7071
 -0.8165 -0.5774 -0.0000
 -0.4082 0.5774 0.7071
R=
 -2.4495 -2.4495 -2.0412
    0 1.7321 1.1547
    0
          0 0.7071
>> v = eig(Q)
v =
-0.6392 + 0.7690i
 -0.6392 - 0.7690i
 1.0000 + 0.0000i
>> norm(v(2))
ans =
  1.0000
>> norm(v(3))
```

```
ans =
```

1

Both the second and third eigenvalues are equal to 1.

Exercise 5.6

```
a)
```

B =

- 1 75
- 1 100
- 1 128
- 1 159
- 1 195

d =

- 15
- 23
- 26
- 34
- 38

$$>> [Q, R] = qr(B, 0)$$

Q =

- -0.4472 -0.5950
- -0.4472 -0.3313
- -0.4472 -0.0359
- -0.4472 0.2912
- -0.4472 0.6710

R=

0 94.7903

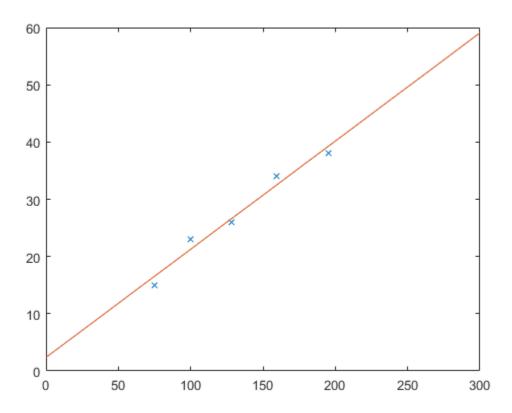
```
>> x = Q(:, 1)
y = Q(:, 2)
x =
 -0.4472
 -0.4472
 -0.4472
 -0.4472
 -0.4472
y =
 -0.5950
 -0.3313
 -0.0359
  0.2912
  0.6710
>> v = dot(x,d)*x + dot(y,d)*y
y =
  16.5379
 21.2640
 26.5572
 32.4176
 39.2232
b)
>> c = B\v
c =
  2.3596
  0.1890
>> B*c - v
```

ans =

1.0e-14 *

```
0.7105
 -0.7105
 -0.3553
     0
     0
>> cl = Iscov(B, d, eye(5))
cl =
  2.3596
  0.1890
The MATLAB answer is the same as our calculated answer.
Exercise 5.7
a)
Because cl = [2.3596; 0.1890], we know that b = 2.3596 and m = 0.1890.
So, we get the equation y = 0.1890x + 2.3596
b)
>> y = 0.1890*35 + 2.3596
y =
  8.9746
>> y = 0.1890*170 + 2.3596
y =
 34.4896
>> y = 0.1890*290 + 2.3596
y =
 57.1696
c)
x = B(:,2);
y = d;
t = 0:1:300;
```

z = polyval([c(2);c(1)],t);;plot(x,y,'x',t,z)



As we can see from the graph, our line is a fairly decent approximation of the data set.