

CS 4220: Some Review Problems (Solutions/Hints)

1. Given data $(t_1, y_1), \dots, (t_m, y_m)$, show how `fminbnd` can be used to solve the problem of minimizing α_1 , α_2 and λ so that

$$\phi(a, b, \lambda) = \sum_{i=1}^m (\alpha_1 + \alpha_2 e^{\lambda t_i} - y_i)^2; = \|A(\lambda)\alpha - y\|_2^2$$

is minimum?. What is $A(\lambda)$? What is the best choice of α given λ ? What is the function passed to `fminbnd`?

```
% One possibility...
function z = MyF(lambda,t,y)
m = length(y);
A = [ ones(m,1) exp(lambda*t) ];
[Q,R] = qr(A);
% The min value of norm(A*alpha - y)*2 is norm(Q2'*y)^2 where
% Q2 = Q(:,3:m).
z = norm(Q(:,3:m)'*y)^2

% Would need to supply a search interval to fminbnd.

% is it possible to compute z using QR(A,0) ?
```

2. Assume the availability of the following function:

```
function [Q,R] = Update(Q0,R0,u,v)
% Q0 is m-by-m orthogonal, R0 is m-by-n upper triangular, u is m-by-1, v is n-by-1
% Q is m-by-m orthogonal, R is m-by-n upper triangular, Q*R = Q0*R0 + u*v'
```

- (a) How would you solve $\min \|Ax - b\|$ and $\min \|\tilde{A}x - b\|$ given that \tilde{A} is defined by $\tilde{A}(:,k) = A(:,k) - \text{sum}(A(:,k))$.
- (b) How would you solve $\min \|Ax - b\|$ and $\min \|\tilde{A}x - b\|$ given that \tilde{A} is A with a new row appended to the bottom.
- (c) How would you solve $\min \|Ax - b\|$ and $\min \|\tilde{A}x - b\|$ given that \tilde{A} is A with its first column replaced by a given vector.

```
% (a)
[m,n] = size(A);
[Q0,R0] = qr(A);
x = R0(1:n,1:n)\(Q0(:,1:n)'*b)
v = sum(A)'; % i.e., v(j) = sum(A(:,j)), j=1:n
u = -ones(m,1)
[Q,R] = update(Q0,R0,u,v);
xtilde = R(1:n,1:n)\(Q(:,1:n)'*b)
```

```

% (b)
[m,n] = size(A)
% adding a zero row to A and zero to b does not change the solution to the LS prob...
A = [A; zeros(1,n)]; b = [b;0];
[Q0,R0] = qr(A);
x = R0(1:n,1:n)\(Q0(:,1:n)'*b)
u = [zeros(m,1);1] % The last column of eye(m+1,m+1)
% Assume that v' is the row to be appended
[Q,R] = Update(Q0,R0,u,v) % Atilde = [A;v']
xtilde = R(1:n,1:n)\(Q(:,1:n)'*b)

```

```

% (c)
% Atilde = A + (newCol - A(:,1)e1' where e1 first col of eye(n,n)
[m,n] = size(A);
[Q0,R0] = qr(A);
x = R0(1:n,1:n)\(Q0(:,1:n)'*b);
% Assume newCol is the new column 1..
u = newCol - A(:,1);
v = [1;zeros(n-1,1)]
[Q,R] = Update(Q0,R0,u,v)
xtilde = R(1:n,1:n)\(Q(:,1:n)'*b)

```

3, $M \in \mathbb{R}^{n \times n}$ has the property that the first $n - 1$ columns of $M - I_n$ are independent. How would you compute a nonzero vector x so $Mx = x$ given that $P(M - I) = LU$ is on tap?

```

% Assume M - I is singular. (FORGOT TO MENTION THIS!)
[L,U,P] = lu(M - eye(n,n));
% P(M - I) = LU. Since M-I is singular, U must be singular.
% So if we can find a nonzero x so  $Ux = 0$ , it follows that  $P(M-I)x = 0$ ,
% and so  $(M-I)x = 0$ , i.e.,  $Mx = x$ .
% Since the first n-1 columns of M-I are independent, the same can
% be said about the first n-1 columns of  $\text{inv}(L)P(M-I) = U$ . So we must
% have  $U(n,n) = 0$ . Set  $x(n) = -1$ . From
%
%           U(1:n-1,1:n-1) U(1:n-1,n)    x(1:n-1)    0
%           *               =
%           0               0       -1           0

```

```

x = zeros(n,1); x(n) = -1; x(1:n-1) = U(1:n-1,1:n-1)\U(1:n-1,n)

```

4. $G \in \mathbb{R}^{n \times n}$ is a 0-1 matrix in sparse format. Define the matrix $B \in \mathbb{R}^{n \times n}$ by

$$B(:,k) = \begin{cases} G(:,k)/\text{sum}(G(:,k)) & \text{if } G(:,k) \neq 0 \\ \text{ones}(n,1)/n & \text{otherwise} \end{cases}$$

Let $0 < \rho < 1$ be a scalar. How would you apply the power method to compute the dominant eigenvector for the matrix $A = \rho B + (1 - \rho) * \text{ones}(n,n)/n$. Show carefully how you would organize the matrix-vector products.

```

% We say v is a probability vector if v(i) >= 0 all i, and sum(v) = 1.
% Power method step:
%   x = A*x ; x = x/norm(x,1)
% We use the 1-norm because if x is a probability vector then Ax is a

```

```

% probability vector.
% Let e = ones(n,1)
%
% Ax = rho*B*x + (1-rho)*(ones(n,n)/n)*x
%     = rho*B*x + ((1-rho)/n)*e'*x
%     = rho*B*x + ((1-rho)/n)*e
%     = rho*(G(:,ip)*(x(ip)./d(ip)) + sum(x(iz))*e/n) + ((1-rho)/n)*e
-
% where
d = sum(G); iz = d==0; ip = d>0;

% Precompute these vectors...
dtilde = d(ip)/rho;
etilde1 = (rho/n)*ones(n,1);
etilde2 = ((1-rho)/n)*ones(n,1)

% How to compute y = A*x
y = G(:,ip)*(x(ip)./dtilde) + sum(x(iz))*etilde1 + etilde2

```

5. $Q \in \mathbb{R}^{n \times n}$ is orthogonal with a unique eigenvalue equal to one. What happens if the power method is applied to the matrix $A = I + (Q + Q^T)/2$? Hint: Draw a picture of A 's eigenvalues.

```

%All the eigenvalues of Q are on the unit circle. (Take 2-norms in Qx = lambda*x.)
% If Qx = lambda*x then
% Ax = x + .5*(Qx + Q'*x) = x + .5*(lambda*x + 1/lambda)*x
%     = x + .5*(lambda + conj(lambda))*x
%     = (1 + real(lambda))*x
% so all the eigenvalues of the symmetric matrix A are in the interval [0,2]
% The eigenvalue 2 is unique. So the power method finds an x so Qx = x.

```

6. (a) How would you minimize $\|Ax\|_2$ subject to the constraint that $x_1 = 1$? (b) Given $t \in \mathbb{R}^m$, how would you determine a scalar τ so that

$$\sum_{k=1}^m [\sin(t_i - \tau)]^2$$

is minimized? Hint: Trig identities and SVD.

```

% (a)
% min norm(Ax) = min norm(A(:,1) + A(:,2:n)*x(2:n)))
xopt = [1; -A(:,2:n)\A(:,1)]

% (b)
% Remembering that t is a column vector,
% The summation = norm(sin(t - tau))^2
%               = norm(sin(t)*cos(tau) - cos(t)*sin(tau))^2
%               = norm([sin(t) -cos(t)]*[cos(tau);sin(tau)])^2
% We want a unit vector v = [cos(tau);sin(tau)] so that
% norm(Av) = min where A = [sin(t) -cos(t)], an m-by-2 matrix.
% That is an SVD problem..
[U,S,V] = svd([sin(t) -cos(t)])
v = V(:,2)
tau = atan2(v(2),v(1))

```

7. What does a step of steepest-descent with exact line search look like when applied to the minimization of $\phi(x) = x^T A x / x^T x$ where $A \in \mathbb{R}^{n \times n}$ is symmetric?

```
% The gradient...
g = (2(x'*x)*A*x - 2(x'*A*x)x)/(x'*x)^2 = 2(A - phi(x)I)x;
% Exact line search requires finding mu so that f(mu) = phi(x - mu*g) is minimum
% Solve f'(mu_opt) = 0 for the optimizing mu
```

8. Newton's method is applied to find a , b , and c so that $(x-a)(x-b)(x-c) = x^3 + 5x^2 - 3x + 4$. Describe the linear system that must be solved each step.

```
% (x-a)(x-b)(x-c) = (x^2 -(a+b)x +ab)(x-c)
%                  = x^3 - x^2(a+b+c) +(ab + bc + ca)x -abc
% Find a zero of
%                  a+b+c+5
%   F(a,b,c) =    ab+bc+ca+3
%                  abc+4
%
%                  1      1      1
%   Jacobian =    b+c   a+c   a+b
%                  bc    ac    ab
```

9. How could Householder tridiagonalization be used to solve the linear system $(A^2 - 5A + 6I)x = b$ where $A \in \mathbb{R}^{n \times n}$ is symmetric?

```
% Solve (A-2I)(A-3I)x = b.
% Solve (A-2I)y = b for y, and (A-3I)x = y for x.
% So, Compute the tridiagonalization A = QTQ' and solve
%   Q(T-2I)Q'y = b for y: y = Q*((T-2I)\(Q'*b))
%   Q(T-3I)Q'x = y for x: x = Q*((T-3I)\(Q'*y))
% Note that T-2I and T-3I are tridiagonal.
```

10. Suppose $A, B \in \mathbb{R}^{m \times n}$ and $m < n/2$. How could you find $x \in \mathbb{R}^n$ so that $Ax = b$ and $Bx = c$ where $b, c \in \mathbb{R}^m$ are given?

```
% We must solve the 2m-by-n underdetermined system [A;B]x = [b;c]
M = [A;B]; d = [b;c];
% Solve Mx = d...
[Q,R] = qr(M')
% Mx = d implies (R'Q'x = d for x
z = R(1:2m,1:2m)'\d
x = Q(:,1:2m)*z
```