Appendix

```
clc
close all
sig = 0.15;
m1 = 0; m2 = 0; m0 = [m1 m2]';
Y = @(t,m) (m(1).*exp(m(2).*t));
t = [1 \ 2 \ 4 \ 5 \ 8]';
y = [3.2939 \ 4.2699 \ 7.1749 \ 9.3008 \ 20.259]';
max iter = 1e6;
tol = 1e-6;
warning('off', 'all')
%generate roughening matrices
n =2;
L0 = eye(n);
N0.1 Report the resulting parameter estimates, the number of iterations it took to converge until the
norm of the change in parameter estimates was less than 10-6, and calculate the x2obs and p-
values.
J = @(t,m) [(exp(m(2)*t)), (m(1)*t.*exp(m(2)*t))]; %exact jacobian
[U,sm,X,V,W] = cgsvd(J(t,mo),L0);
%reatools
[req_corner,rho,eta,req_param] = I_curve(U,sm,v,'Tikh');
alpha = rea corner:
disp(['The resulting value of alpha = ',num2str(alpha)]);
for k = 1:max iter
  dm = inv(J(t,mo))^*J(t,mo) + (alpha^2).*L0)^*J(t,mo)^*(y-Y(t,mo));
  m = mo + dm;
  if norm(m-mo) < tol
     fprintf('Number of iterations taken = %d',k);
     break
  end
  mo = m;
disp(['Resulting parameter estimates are [',num2str(m'),']']);
%chi-square
chi_s = 0;
for i = 1:length(t)
  chi_s = chi_s + ((Y(t(i),m)-y(i)))^2;
end
disp(['chi-square obs = ',num2str(chi_s)])
The value of the is much less than the expected value (m-n = 3), hence we reject the null hypothesis.
% pvalue
```

```
p = 1 - chi2cdf(chi_s,5);
disp(['pvalue = ',num2str(p)])
N0.2
m1 = 0; m2 = 0; m0 = [m1 m2]';
K = 10;
rho = []; eta = [];
rrh=[]; al = [];
est = \Pi:
for alpha = logspace(-2,2,20)
  for k = 1:K
   dm = -inv(J(t,mo))*J(t,mo) + (alpha^2).*L0*L0)*(J(t,mo))*(Y(t,mo) - y) + (alpha^2).*L0*L0*mo);
    m = mo + dm;
   if norm(m-mo) < tol
     est = [est m];
     break
    end
   mo = m;
  end
  al = [al \ alpha];
  rho = [rho norm(Y(t,m) - y)];
  res = [rho norm(Y(t,m) - y)^2];
  rrh = [rrh norm(Y(t,m) - y)^2 + norm(m)^2];
  eta = [eta norm(L0*m)];
end
fprintf('Number of iterations taken = %d',k);
M = [est(k-1); est(k)];
disp(['Resulting parameter estimates are [',num2str(M'),']']);
For the final value for k, plot the L-curve and discuss its shape.
figure(1)
loglog(rho,eta,'ok-')
ylabel('solution semi-norm | Lm | 2')
xlabel('residual norm || Gm - y ||_2')
figure(3)
subplot(2,1,1)
plot(al,res(1:end-1),'ok-')
xlabel('\alpha')
ylabel('|| Gm - y ||_2^2 ')
title('Residual norm')
subplot(2,1,2)
plot(al,rrh,'ok-')
xlabel('\alpha')
ylabel('|| Gm - y || 2^2 + ||m|| 2^2')
title('Regularized residual norm')
```

```
N0.3 Occam's inversion algorithm
m1 = 0; m2 = 0; m0 = [m1 m2]';
M = []; R alp = [];
K=15; m = 5; n = 2;
delta = (m-n)*sig^2;
alphas = logspace(-2,0.5,K);
for k = 1:K
  chi = \Pi:
  dcap = y - Y(t,mo) + J(t,mo)*mo;
  for alpha = logspace(-2,0.5,K)
   mk = inv(J(t,mo)' * J(t,mo) + (alpha^2) .* L0' * L0) * J(t,mo)' * dcap ;
   chii = norm(Y(t,mk) - y)^2;
   chi = [chi chii];
  end
  [Y1,idx] = min(chi); R_alp = [R_alp Y1];
  if (Y1 > delta^2)
     alpha = alphas(idx(1));
  else
     idx = find(chi \ll delta^2);
     alpha = alphas(max(idx));
  end
  mo = inv(J(t,mo)) * J(t,mo) + (alpha^2) .* L0 * L0 * J(t,mo) * dcap;
  M = [M mo];
end
Resulting_estimates = M
Resulting_alpha = R_alp
```