

```
% gda05_16
%
% total distribution pT(m1,d1) and the projected
% distribution p(m) = (integral) pT(m1,d1) d(d1)
% highlighting the possibility that the maximum
% liklihood points of the two distributions are
% at two different values of m1
```

```
% Supports Figure 5.16.
```

```
clear all;
```

```
% d1 variable
```

```
Nd1 = 101;
dlmin = 0;
dlmax = 5.0;
Dd1 = (dlmax-dlmin)/(Nd1-1);
d1 = dlmin + Dd1*[0:Nd1-1]';
```

```
% m1 variable
```

```
Nm1 = 101;
mlmin = 0;
mlmax = 5.0;
Dm1 = (mlmax-mlmin)/(Nm1-1);
m1 = mlmin + Dm1*[0:Nm1-1]';
```

```
% distribution P1 = pA
```

```
P1=zeros(Nd1,Nm1);
dlbar = 2.25;
mlbar = 2.08;
bar = [dlbar, mlbar]';
sd1 = 0.5;
sm1 = 1;
C1 = diag( [sd1^2, sm1^2]' );
CI1 = inv(C1);
DC1 = det(C1);
% note not normaalized, so max is unity
for i=[1:Nm1]
for j=[1:Nd1]
    x1=[d1(i), m1(j)]' - bar;
    P1(i,j) = exp( -0.5 * x1'*CI1*x1 );
end
end
```

```
% axis for parametric curve
```

```
Ns = 51;
smin = 0;
smax = 5.0;
Ds = (smax-smin)/(Ns-1);
s = smin + Ds*[0:Ns-1]';
```

```
% parameric curve
```

```
dp = 1+s-2*(s/smax).^2;
mp = s;
```

```
% P2 = Pg
```

```
P2 = zeros(Nd1,Nm1);
sg = 0.35;
sg2 = sg^2;
```

```

for i=[1:Nm1]
for j=[1:Nd1]
    r2 = (d1(i)-dp).^2 + (m1(j)-mp).^2;
    r2min=min(r2);
    P2(i,j) = exp( -r2min/sg2 );
end
end

% P3 = pT with pT=pA pg
P3 = P1.*P2;
[tmp, itmp] = max(P3);
[P3max, Pj] = max(tmp);
Pi=itmp(Pj);
Pmaxm1 = m1min+Dm1*(Pj-1);
Pmaxd1 = d1min+Dd1*(Pi-1);

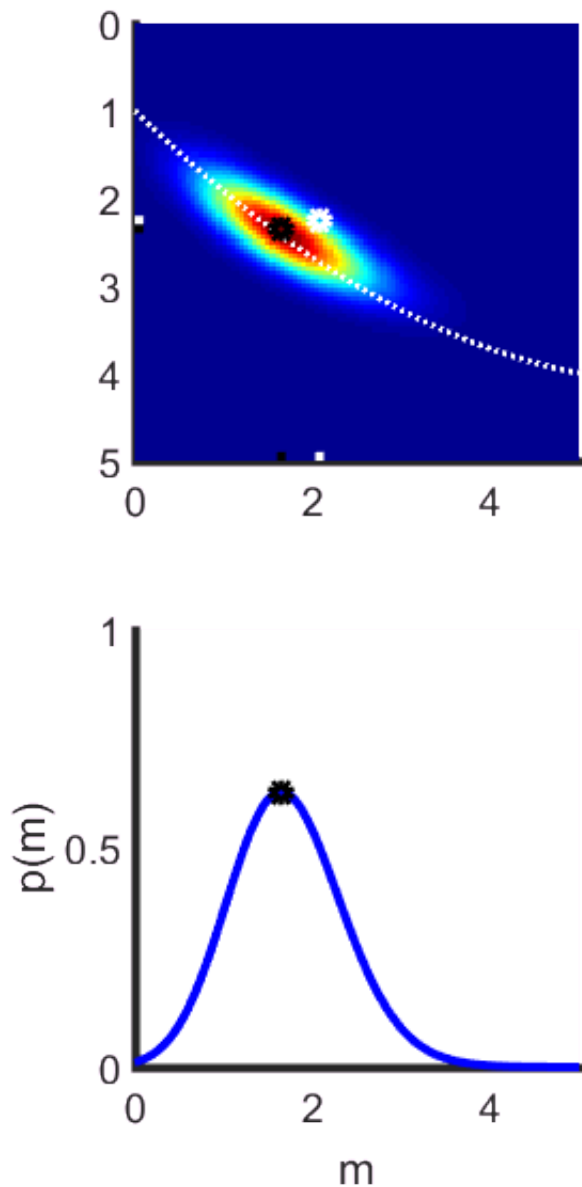
% find maximum likelihood point
Pm = sum(P3);
[Pmmax, Pmi] = max(Pm);
Pmm = m1min+Dm1*(Pmi-1);
norm=Dm1*sum(Pm);
Pm = Pm/norm;

figure(1);
set(gcf,'pos',[10, 10, 300, 600] );
clf;

% plot pT(d1,m1)
subplot(2,1,1);
set(gca,'LineWidth',3);
set(gca,'FontSize',14);
colormap('jet');
hold on;
axis( [d1min, d1max, m1min, m1max] );
axis ij;
imagesc( [d1min, d1max], [m1min, m1max], P3 );
plot( mp, dp, 'w:', 'LineWidth', 2 );
plot( mlbar, dlbar, 'wo', 'LineWidth', 3 );
plot( Pmaxm1, Pmaxd1, 'ko', 'LineWidth', 3 );
plot( [mlbar, mlbar], [d1max, d1max-0.1], 'w-', 'LineWidth', 3 );
plot( [m1min, m1min+0.1], [dlbar, dlbar], 'w-', 'LineWidth', 3 );
plot( [Pmaxm1, Pmaxm1], [d1max, d1max-0.1], 'k-', 'LineWidth', 3 );
plot( [m1min, m1min+0.1], [Pmaxd1, Pmaxd1], 'k-', 'LineWidth', 3 );

% plot p(m1)
subplot(2,1,2);
set(gca,'LineWidth',3);
set(gca,'FontSize',14);
colormap('jet');
hold on;
axis( [m1min, m1max, 0, 1] );
axis xy;
plot( m1, Pm, 'b-', 'LineWidth', 3 );
plot( Pmm, max(Pm), 'ko', 'LineWidth', 3 );
xlabel('m');
ylabel('p(m)');

```



% Figure 5.16 (A) The joint probability density function $p_T(m, d)$ can be considered the solution to the inverse problem. Its maximum likelihood point (black circle) gives an estimate of the parameter m_{est} and a prediction of the data d_{pre} . (B) The function $p_T(m, d)$ is projected onto the m -axis, by integrating over d , to form the probability density function $p(m)$ of the model parameter m irrespective of the datum. This function also has a maximum likelihood point m_{est}' which in general can be different than m_{est} . The distinction points out the difficulty of defining a unique solution to an inverse problem. MatLab script gda05_16.