

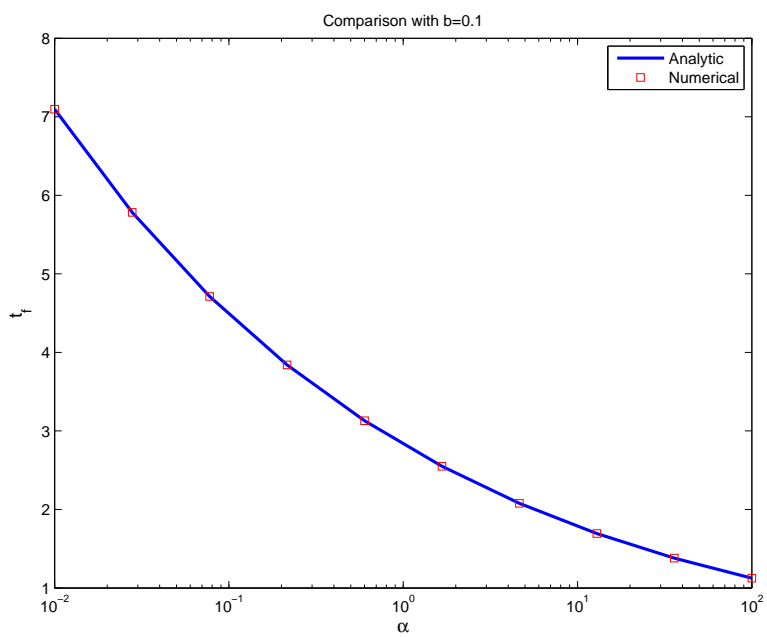
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16.323 Principles of Optimal Control
Spring 2008

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16.323 Lecture 7

Numerical Solution in Matlab



- Performance

$$J = \int_{t_0}^{t_f} (u - x)^2 dt$$

with dynamics $\dot{x} = u$ and BC $t_0 = 0$, $x_0 = 1$, $t_f = 1$.

– So this is a fixed final time, free final state problem.

- Form Hamiltonian

$$H = (u - x)^2 + pu$$

- Necessary conditions become:

$$\dot{x} = u \tag{7.25}$$

$$\dot{p} = -2(u - x)(-1) \tag{7.26}$$

$$0 = 2(u - x) + p \tag{7.27}$$

with BC that $p(t_f) = 0$.

- Rearrange to get

$$\dot{p} = -p \tag{7.28}$$

$$\Rightarrow p(t) = c_1 e^{-t} \tag{7.29}$$

But now impose BC to get

$$p(t) = 0 \tag{7.30}$$

- This implies that $u = x$ is the optimal solution, and the closed-loop dynamics are

$$\dot{x} = x$$

with solution $x(t) = e^t$.

– Clearly this would be an unstable response on a longer timescale, but given the cost and the short time horizon, this control is the best you can do.

- Consider ship that has to travel through a region of strong currents. The ship is assumed to have constant speed V but its heading θ can be varied. The current is assumed to be in the y direction with speed of w .
- The motion of the boat is then given by the dynamics

$$\dot{x} = V \cos \theta \quad (7.31)$$

$$\dot{y} = V \sin \theta + w \quad (7.32)$$

- The goal is to minimize time, the performance index is

$$J = \int_0^{t_f} 1 dt = t_f$$

– with BC $x_0 = y_0 = 0, x_f = 1, y_f = 0$

– Final time is unspecified.

- Define costate $\mathbf{p} = [p_1 \ p_2]^T$, and in this case the Hamiltonian is

$$H = 1 + p_1(V \cos \theta) + p_2(V \sin \theta + w)$$

- Now use the necessary conditions to get ($\dot{\mathbf{p}} = -H_{\mathbf{x}}^T$)

$$\dot{p}_1 = -\frac{\partial H}{\partial x} = 0 \quad \rightarrow p_1 = c_1 \quad (7.33)$$

$$\dot{p}_2 = -\frac{\partial H}{\partial y} = 0 \quad \rightarrow p_2 = c_2 \quad (7.34)$$

- Control input $\theta(t)$ is unconstrained, so have ($H_{\mathbf{u}} = 0$)

$$\frac{\partial H}{\partial u} = -p_1 V \sin \theta + p_2 V \cos \theta = 0 \quad (7.35)$$

which gives the control law

$$\tan \theta = \frac{p_2}{p_1} = \frac{-p_2}{-p_1} \quad (7.36)$$

– Since p_1 and p_2 are constants, then $\theta(t)$ is also a constant.

- Optimal control is constant, so can integrate the state equations:

$$x = Vt \cos \theta \quad (7.37)$$

$$y = Vt(\sin \theta + w) \quad (7.38)$$

– Now impose the BC to get $x(t_f) = 1$, $y(t_f) = 0$ to get

$$t_f = \frac{1}{V \cos \theta} \quad \sin \theta = -\frac{w}{V}$$

- Rearrange to get

$$\cos \theta = \frac{\sqrt{V^2 - w^2}}{V}$$

which gives that

$$t_f = \frac{1}{\sqrt{V^2 - w^2}} \quad \theta = -\arcsin \frac{w}{V}$$

– Does this make sense?

- Most of the problems considered so far have been simple. Things get more complicated by the need to solve a two-point boundary value problem when the dynamics are nonlinear.
- Numerous solution techniques exist, including shooting methods¹³ and collocation
 - Will discuss the details on these later, but for now, let us look at how to solve these use existing codes

- Matlab code called BVP4C exists that is part of the standard package¹⁴
 - Solves problems of a “standard form”:

$$\dot{\mathbf{y}} = \mathbf{f}(\mathbf{y}, t, \mathbf{p}) \quad a \leq t \leq b$$

where \mathbf{y} are the variables of interest, and \mathbf{p} are extra variables in the problem that can also be optimized

- Where the system is subject to the boundary conditions:

$$\mathbf{g}(\mathbf{y}(a), \mathbf{y}(b)) = 0$$

- The solution is an approximation $\mathbf{S}(t)$ which is a continuous function that is a cubic polynomial on sub-intervals $[t_n, t_{n+1}]$ of a mesh

$$a = t_0 < t_1 < \dots < t_{n-1} < t_n = b$$

- This approximation satisfies the boundary conditions, so that:

$$\mathbf{g}(\mathbf{S}(a), \mathbf{S}(b)) = 0$$

- And it satisfies the differential equations (collocates) at both ends and the mid-point of each subinterval:

$$\begin{aligned} \dot{\mathbf{S}}(t_n) &= \mathbf{f}(\mathbf{S}(t_n), t_n) \\ \dot{\mathbf{S}}((t_n + t_{n+1})/2) &= \mathbf{f}(\mathbf{S}((t_n + t_{n+1})/2), (t_n + t_{n+1})/2) \\ \dot{\mathbf{S}}(t_{n+1}) &= \mathbf{f}(\mathbf{S}(t_{n+1}), t_{n+1}) \end{aligned}$$

¹³Online reference

¹⁴Matlab help and BVP4C Tutorial

- Now constrain continuity in the solution at the mesh points \Rightarrow converts problem to a series of nonlinear algebraic equations in the unknowns
 - Becomes a “root finding problem” that can be solved iteratively (Simpson’s method).
- Inputs to BVP4C are functions that evaluate the differential equation $\dot{\mathbf{y}} = \mathbf{f}(\mathbf{y}, t)$ and the residual of the boundary condition (e.g. $y_1(a) = 1$, $y_2(a) = y_1(b)$, and $y_3(b) = 0$):

```
function res = bvpbc(ya, yb)
    res = [ ya(1) - 1
            ya(2) - yb(1)
            yb(3)];
```

- Redo example on page 4–15 using numerical techniques
 - Finite time LQR problem with $t_f = 10$

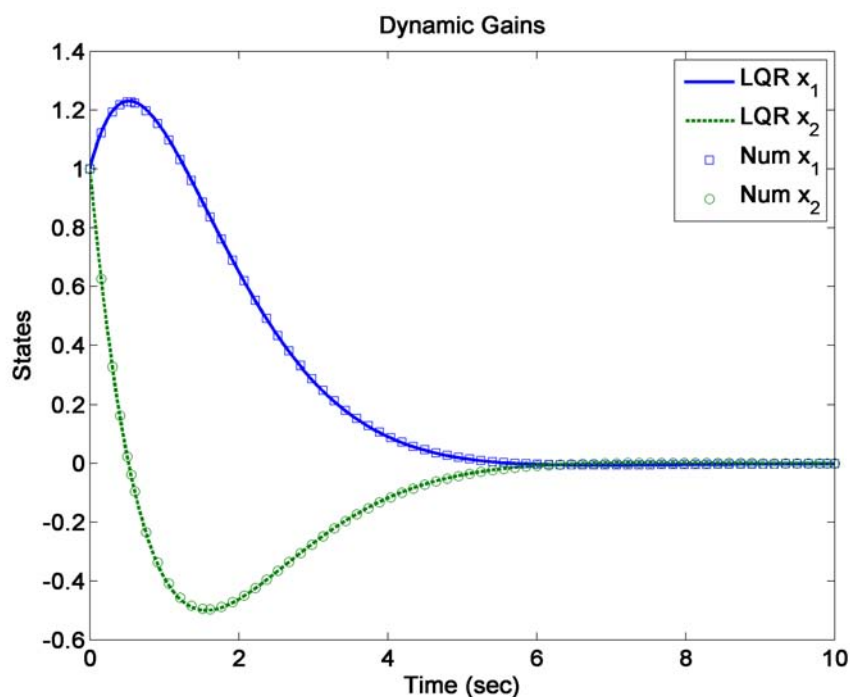


Figure 7.1: Results suggest a good comparison with the dynamic LQR result

TPBVP for LQR

```

1 function m = TPBVPlqr(p1,p2,p3)
2 global A B x0 Rxx Ruu Ptf
3 t_f=10;x0=[1 1]';
4 Rxx=p1;Ruu=p2;Ptf=p3;
5 solinit = bvpinit(linspace(0,t_f),@TPBVPlqrinit);
6 sol = bvp4c(@TPBVPlqrode,@TPBVPlqrbcb,solinit);
7 time = sol.x;
8 state = sol.y([1 2],:);
9 adjoint = sol.y([3 4],:);
10 control = -inv(Ruu)*B'*sol.y([3 4],:);
11 m(1,:) = time;m([2 3],:) = state;m([4 5],:) = adjoint;m(6,:) = control;
12 %-----
13 function dydt=TPBVPlqrode(t,y)
14 global A B x0 Rxx Ruu Ptf
15 dydt=[ A -B/Ruu*B'; -Rxx -A']*y;
16 %-----
17 function res=TPBVPlqrbcb(ya,yb)
18 global A B x0 Rxx Ruu Ptf
19 res=[ya(1) - x0(1);ya(2)-x0(2);yb(3:4)-Ptf*yb(1:2)];
20 %-----
21 function v=TPBVPlqrinit(t)
22 global A B x0 b alp
23 v=[x0;1;0];
24 return
25
26 % 16.323 Spring 2007
27 % Jonathan How
28 % redo LQR example on page 4-15 using numerical approaches
29 clear all;close all;
30 set(0, 'DefaultAxesFontSize', 14, 'DefaultAxesFontWeight','demi')
31 set(0, 'DefaultTextFontSize', 14, 'DefaultTextFontWeight','demi')
32 %
33 global A B
34 Ptf=[0 0;0 4];Rxx=[1 0;0 0];Ruu=1;A=[0 1;0 -1];B=[0 1]';
35 tf=10;dt=.01;time=[0:dt:tf];
36 m=TPBVPlqr(Rxx,Ruu,Ptf); % numerical result
37
38 % integrate the P backwards for LQR result
39 P=zeros(2,2,length(time));K=zeros(1,2,length(time));
40 Pcurr=Ptf;
41 for kk=0:length(time)-1
42     P(:,:,length(time)-kk)=Pcurr;
43     K(:,:,length(time)-kk)=inv(Ruu)*B'*Pcurr;
44     Pdot=-Pcurr*A-A'*Pcurr-Rxx+Pcurr*B*inv(Ruu)*B'*Pcurr;
45     Pcurr=Pcurr-dt*Pdot;
46 end
47
48 % simulate the state
49 x1=zeros(2,1,length(time));xcurr1=[1 1]';
50 for kk=1:length(time)-1
51     x1(:,,kk)=xcurr1;
52     xdot1=(A-B*K(:,:,kk))*x1(:,,kk);
53     xcurr1=xcurr1+xdot1*dt;
54 end
55
56 figure(3);clf
57 plot(time,squeeze(x1(1,1,:)),time,squeeze(x1(2,1,:)),'--','LineWidth',2),
58 xlabel('Time (sec)');ylabel('States');title('Dynamic Gains')
59 hold on;plot(m(1,:),m([2],:),'s',m(1,:),m([3],:),'o');hold off
60 legend('LQR x_1','LQR x_2','Num x_1','Num x_2')
61 print -dpng -r300 numreg2.png

```

- BVP4C sounds good, but this standard form doesn't match many of the problems that we care about
 - In particular, free end time problems are excluded, because the time period is defined to be fixed $t \in [a, b]$
- Can convert our problems of interest into this standard form though using some pretty handy tricks.
 - U. Ascher and R. D. Russell, "Reformulation of Boundary Value Problems into "Standard" Form," *SIAM Review*, Vol. 23, No. 2, 238-254. Apr., 1981.
- Key step is to re-scale time so that $\tau = t/t_f$, then $\tau \in [0, 1]$.
 - Implications of this scaling are that the derivatives must be changed since $d\tau = dt/t_f$
$$\frac{d}{d\tau} = t_f \frac{d}{dt}$$
- Final step is to introduce a dummy state r that corresponds to t_f with the trivial dynamics $\dot{r} = 0$.
 - Now replace all instances of t_f in the necessary/boundary conditions for state r .
 - Optimizer will then just pick an appropriate constant for $r = t_f$

- Recall that our basic set of necessary conditions are, for $t \in [t_0, t_f]$

$$\begin{aligned}\dot{\mathbf{x}} &= \mathbf{a}(\mathbf{x}, \mathbf{u}, t) \\ \dot{\mathbf{p}} &= -H_{\mathbf{x}}^T \\ H_{\mathbf{u}} &= 0\end{aligned}$$

- And we considered various boundary conditions $\mathbf{x}(t_0) = \mathbf{x}_0$, and:
 - If t_f is free: $h_t + g + \mathbf{p}^T \mathbf{a} = h_t + H(t_f) = 0$
 - If $\mathbf{x}_i(t_f)$ is fixed, then $\mathbf{x}_i(t_f) = x_{i_f}$
 - If $\mathbf{x}_i(t_f)$ is free, then $\mathbf{p}_i(t_f) = \frac{\partial h}{\partial x_i}(t_f)$

- Then

$$\dot{\mathbf{x}} = \mathbf{a}(\mathbf{x}, \mathbf{u}, t) \Rightarrow \mathbf{x}' = t_f \mathbf{a}(\mathbf{x}, \mathbf{u}, \tau)$$

and

$$\dot{\mathbf{p}} = -H_{\mathbf{x}}^T \Rightarrow \mathbf{p}' = -t_f H_{\mathbf{x}}^T$$

- Revisit example on page 6-6
- Linear system with performance/time weighting and free end time
 - Necessary conditions are:

$$\begin{aligned}\dot{\mathbf{x}} &= A\mathbf{x} + Bu \\ \dot{\mathbf{p}} &= -A^T \mathbf{p} \\ 0 &= bu + [0 \ 1] \mathbf{p}\end{aligned}$$

with state conditions

$$\begin{aligned}\mathbf{x}_1(0) &= 10 \\ \mathbf{x}_2(0) &= 0 \\ \mathbf{x}_1(t_f) &= 0 \\ \mathbf{x}_2(t_f) &= 0 \\ -0.5bu^2(t_f) + \alpha t_f &= 0\end{aligned}$$

- Define the state of interest $\mathbf{z} = [\mathbf{x}^T \ \mathbf{p}^T \ r]^T$ and note that

$$\begin{aligned}\frac{d\mathbf{z}}{d\tau} &= t_f \frac{d\mathbf{z}}{dt} \\ &= \mathbf{z}_5 \begin{bmatrix} A & -B [0 \ 1] / b & 0 \\ 0 & -A^T & 0 \\ 0 & 0 & 0 \end{bmatrix} \mathbf{z} \\ \Rightarrow \mathbf{z}' &= f(\mathbf{z}) \quad \text{which is nonlinear}\end{aligned}$$

with BC:

$$\begin{aligned}\mathbf{z}_1(0) &= 10 \\ \mathbf{z}_2(0) &= 0 \\ \mathbf{z}_1(1) &= 0 \\ \mathbf{z}_2(1) &= 0 \\ \frac{-0.5}{b} \mathbf{z}_4^2(1) + \alpha \mathbf{z}_5(1) &= 0\end{aligned}$$

- Code given on following pages
 - Note – it is not particularly complicated
 - Solution time/iteration count is a strong function of the initial solution – not a particularly good choice for \mathbf{p} is used here

- Analytic solution gave $t_f = (1800b/\alpha)^{1/5}$
 - Numerical result give close agreement in prediction of the final time

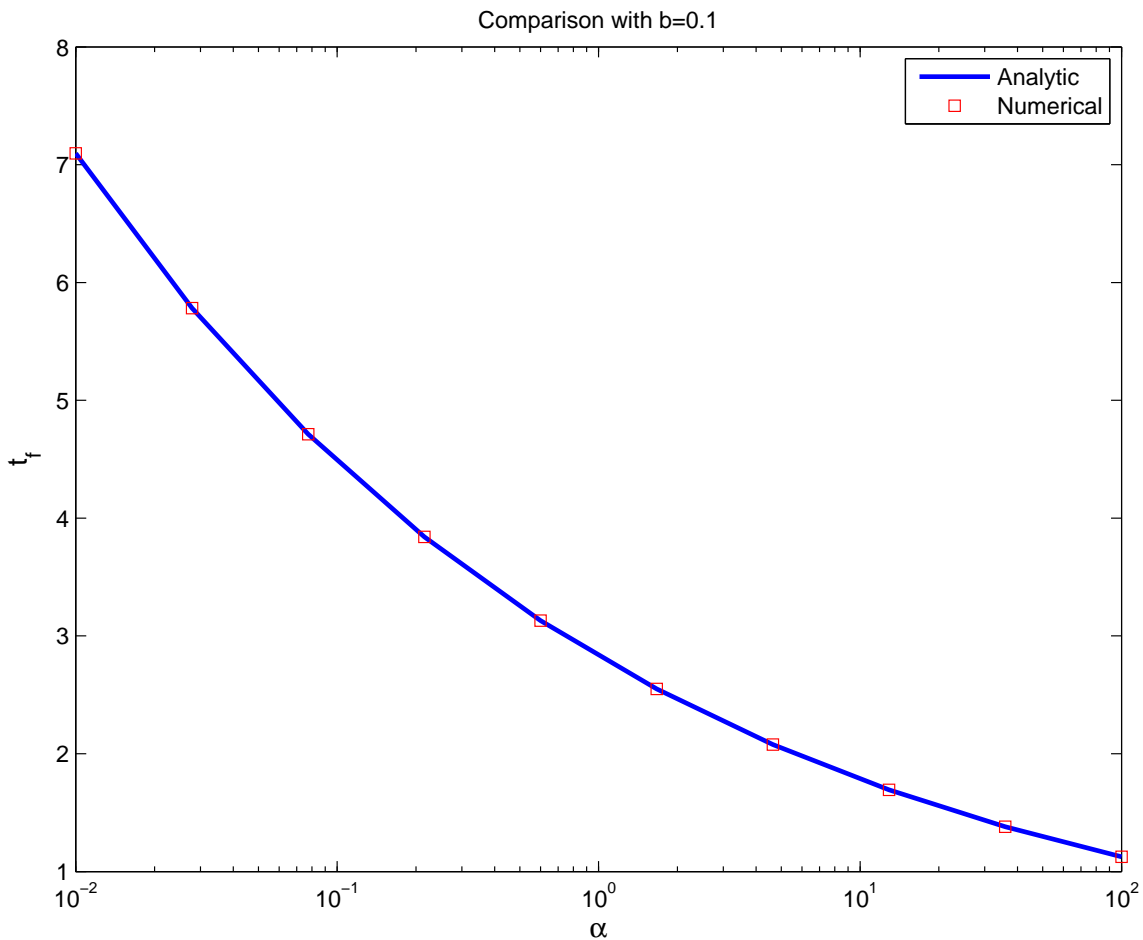


Figure 7.2: Comparison of the predicted completion times for the maneuver

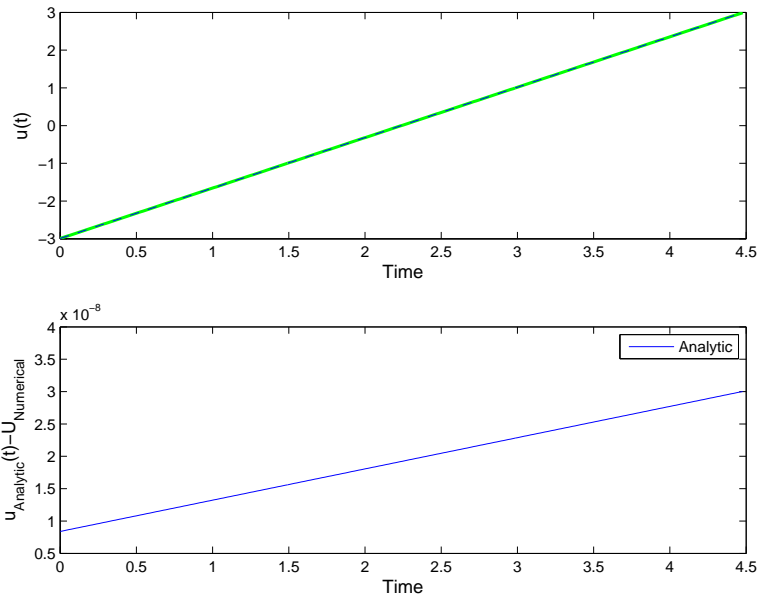


Figure 7.3: Control Inputs

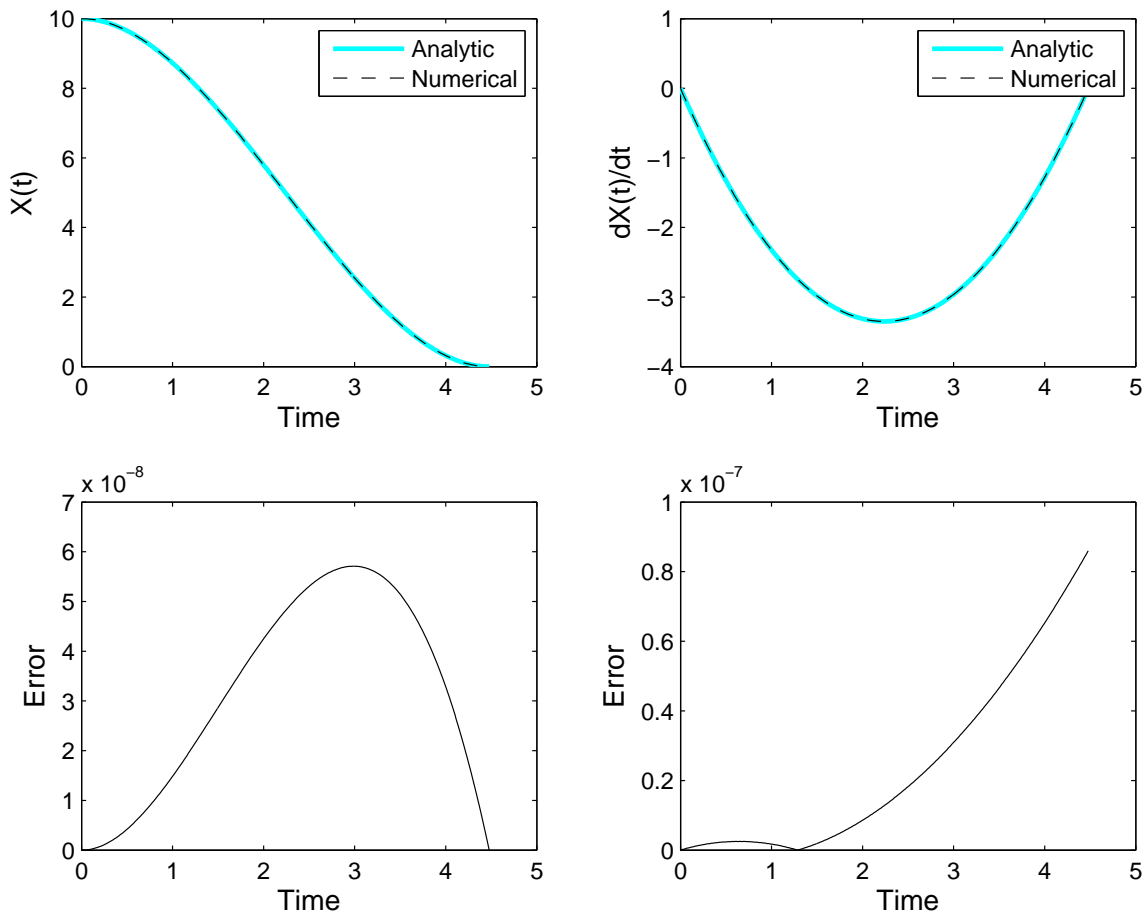


Figure 7.4: State response

 TPBVP

```

1 function m = TPBVP(p1,p2)
2 % 16.323 Spring 2007
3 % Jonathan How
4 %
5 global A B x0 b alp
6
7 A=[0 1;0 0];
8 B=[0 1]';
9 x0=[10 0]';
10 b=p1;
11 alp=p2;
12
13 solinit = bvpinit(linspace(0,1),@TPBVPinit);
14 sol = bvp4c(@TPBVPode,@TPBVPbc,solinit);
15
16 time = sol.y(5)*sol.x;
17 state = sol.y([1 2],:);
18 adjoint = sol.y([3 4],:);
19 control = -(1/b)*sol.y(4,:);
20 m(1,:) = time;
21 m([2 3],:) = state;
22 m([4 5],:) = adjoint;
23 m(6,:) = control;
24
25 %-----
26 function dydt=TPBVPode(t,y)
27 global A B x0 b alp
28 dydt=y(5)*[ A -B*[0 1]/b zeros(2,1); zeros(2,2) -A' zeros(2,1);zeros(1,5)]*y;
29
30 %-----
31 function res=TPBVPbc(ya,yb)
32 global A B x0 b alp
33 res=[ya(1) - x0(1);ya(2)-x0(2);yb(1);yb(2);-0.5*yb(4)^2/b+ alp*yb(5)];
34
35 %-----
36 function v=TPBVPinit(t)
37 global A B x0 b alp
38 v=[x0;1;0;1];
39
40 return
41

```

 TPBVP Main

```

1 % 16.323 Spring 2007
2 % Jonathan How
3 % TPmain.m
4 %
5 b=0.1;
6 %alp=[.05 .1 1 10 20];
7 alp=logspace(-2,2,10);
8 t=[];
9 for alpha=alp
10     m=TPBVP(b,alpha);
11     t=[t;m(1,end)];
12 end
13
14 figure(1);clf
15 semilogx(alp,(1800*b./alp).^0.2,'-', 'LineWidth',2)
16 hold on;semilogx(alp,t,'rs');hold off
17 xlabel('\alpha', 'FontSize',12);ylabel('t_f', 'FontSize',12)
18 legend('Analytic', 'Numerical')
19 title('Comparison with b=0.1')
20 print -depsc -f1 TPBVP1.eps;jpdf('TPBVP1')
21
22 % code from opt1.m on the analytic solution
23 b=0.1;alpha=0.1;
24 m=TPBVP(b,alpha);
25 tf=(1800*b/alpha)^0.2;
26 c1=120*b/tf^3;
27 c2=60*b/tf^2;
28 u=(-c2+c1*m(1,:))/b;
29 A=[0 1;0 0];B=[0 1]';C=eye(2);D=zeros(2,1);G=ss(A,B,C,D);X0=[10 0]';
30 [y3,t3]=lsim(G,u,m(1,:),X0);
31
32 figure(2);clf
33 subplot(211)
34 plot(m(1,:),u,'g-', 'LineWidth',2);
35 xlabel('Time', 'FontSize',12);ylabel('u(t)', 'FontSize',12)
36 hold on;plot(m(1,:),m(6,:), '--');hold off
37 subplot(212)
38 plot(m(1,:),abs(u-m(6,:)), '-');
39 xlabel('Time', 'FontSize',12)
40 ylabel('u_{Analytic}(t)-U_{Numerical}', 'FontSize',12)
41 legend('Analytic', 'Numerical')
42 print -depsc -f2 TPBVP2.eps;jpdf('TPBVP2')
43
44 figure(3);clf
45 subplot(221)
46 plot(m(1,:),y3(:,1),'c-', 'LineWidth',2);
47 xlabel('Time', 'FontSize',12);ylabel('X(t)', 'FontSize',12)
48 hold on;plot(m(1,:),m([2],:),'k--');hold off
49 legend('Analytic', 'Numerical')
50 subplot(222)
51 plot(m(1,:),y3(:,2),'c-', 'LineWidth',2);
52 xlabel('Time', 'FontSize',12);ylabel('dX(t)/dt', 'FontSize',12)
53 hold on;plot(m(1,:),m([3],:),'k--');hold off
54 legend('Analytic', 'Numerical')
55 subplot(223)
56 plot(m(1,:),abs(y3(:,1)-m(2,:)),'k-');
57 xlabel('Time', 'FontSize',12);ylabel('Error', 'FontSize',12)
58 subplot(224)
59 plot(m(1,:),abs(y3(:,2)-m(3,:)),'k-');
60 xlabel('Time', 'FontSize',12);ylabel('Error', 'FontSize',12)
61 print -depsc -f3 TPBVP3.eps;jpdf('TPBVP3')

```

- Simplified dynamics of a UAV flying in a horizontal plane can be modeled as:

$$\begin{aligned}\dot{x}(t) &= V \cos \theta(t) \\ \dot{y}(t) &= V \sin \theta(t) + w\end{aligned}$$

where $\theta(t)$ is the heading angle (control input) with respect to the x axis, V is the speed.

- Objective: fly from point A to B in minimum time:

$$\min J = \int_0^{t_f} (1) dt$$

where t_f is free.

– Initial conditions are:

$$x(0) = x_0 \qquad y(0) = y_0$$

– Final conditions are:

$$x(t_f) = x_1 \qquad y(t_f) = y_1$$

- Apply the standard necessary conditions with

$$H = 1 + p_1 V(\cos \theta(t)) + p_2(V \sin \theta(t) + w)$$

$$\begin{aligned} \dot{\mathbf{x}} &= \mathbf{a}(\mathbf{x}, \mathbf{u}, t) \\ \dot{\mathbf{p}} &= -H_{\mathbf{x}}^T \\ H_{\mathbf{u}} &= 0 \end{aligned}$$

$$\dot{x}(t) = V \cos \theta(t)$$

$$\dot{y}(t) = V \sin \theta(t) + w$$

$$\dot{p}_1(t) = 0$$

$$\dot{p}_2(t) = 0$$

$$0 = -p_1 \sin \theta(t) + p_2 \cos \theta(t)$$

– Then add extra state for the time.

- Since t_f is free, must add terminal condition that $H(t_f) = 0$, which gives a total of 5 conditions (2 initial, 3 terminal).

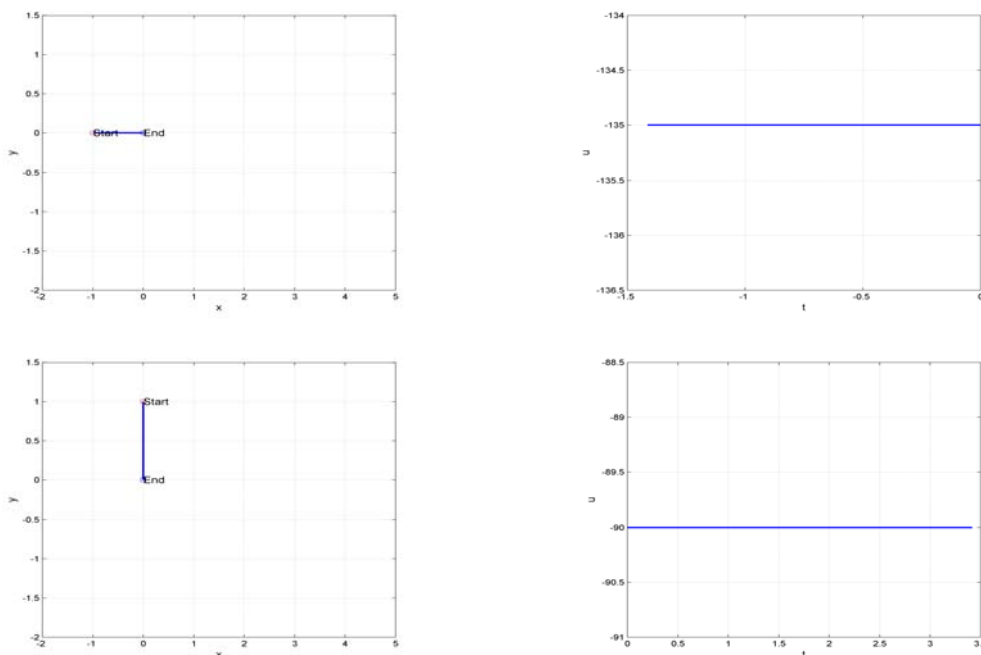


Figure 7.5: Zermelo examples

 TPBVPZermelo

```

1 function m = TPBVPzermelo(p1,p2)
2 global x0 x1 V w
3
4 solinit = bvpinit(linspace(0,1),@TPBVPinit);
5 sol = bvp6c(@TPBVPode,@TPBVPbc,solinit);
6
7 time = sol.y(5)*sol.x;
8 state = sol.y([1 2],:);
9 adjoint = sol.y([3 4],:);
10 control = atan2(-sol.y([4],:),-sol.y([3],:));
11
12 m(1,:) = time;
13 m([2 3],:) = state;
14 m([4 5],:) = adjoint;
15 m(6,:) = control;
16 return
17
18 %-----
19 function dydt=TPBVPode(t,y)
20 global x0 x1 V w
21
22 % x y p1 p2 t
23 % minimizing form
24 sinh=-y(4)/sqrt(y(3)^2+y(4)^2);
25 cosh=-y(3)/sqrt(y(3)^2+y(4)^2);
26
27 dydt=y(5)*[V*cosh ; V*sinh+w;0;0;0];
28 %-----
29 function res=TPBVPbc(ya,yb)
30 global x0 x1 V w
31 % x y p1 p2 t
32 % minimizing form
33 cosh=-yb(3)/sqrt(yb(3)^2+yb(4)^2);
34 sinh=-yb(4)/sqrt(yb(3)^2+yb(4)^2);
35
36 res=[ya(1) - x0(1);ya(2)-x0(2);
37       yb(1) - x1(1);yb(2)-x1(2);
38       1+V*cosh*yb(3)+V*(sinh+w)*yb(4)];
39
40 %-----
41 function v=TPBVPinit(t)
42 global x0 x1 V w
43 %v=[x0;-1;-1;norm(x1-x0)/(V-w)];
44 v=[x0;1;1;norm(x1-x0)/(V-w)];
45 return
46
47 clear all
48 global x0 x1 V w
49 w=1/sqrt(2);
50 x0=[-1 0]';x1=[0 0]';V = 1;
51 mm=TPBVPzermelo;
52
53 figure(1);clf
54 plot(mm(2,:),mm([3],:),'LineWidth',2);axis('square');grid on
55 axis([-2 5 -2 1.5 ])
56 xlabel('x','FontSize',12);ylabel('y','FontSize',12);
57 hold on;
58 plot(x0(1),x0(2),'rs');plot(x1(1),x1(2),'bs');
59 text(x0(1),x0(2),'Start','FontSize',12)
60 text(x1(1),x1(2),'End','FontSize',12)
61 hold off
62
63 figure(2);clf
64 plot(mm(1,:),180/pi*mm([6],:),'LineWidth',2);grid on;axis('square')
65 xlabel('t','FontSize',12);ylabel('u','FontSize',12);
66
67 print -dpng -r300 -f1 BVP_zermelo.png;

```

```
68 print -dpng -r300 -f2 BVP_zermelo2.png;
69
70 clear all
71 global x0 x1 V w
72 w=1/sqrt(2);
73 x0=[0 1]';x1=[0 0]';V = 1;
74 mm=TPBVPzermelo;
75
76 figure(1);clf
77 plot(mm(2,:),mm([3],:),'LineWidth',2);axis('square');grid on
78 axis([-2 5 -2 1.5 ])
79 xlabel('x','FontSize',12);ylabel('y','FontSize',12);
80 hold on;
81 plot(x0(1),x0(2),'rs');plot(x1(1),x1(2),'bs');
82 text(x0(1),x0(2),'Start','FontSize',12)
83 text(x1(1),x1(2),'End','FontSize',12)
84 hold off
85
86 figure(2);clf
87 plot(mm(1,:),180/pi*mm([6],:),'LineWidth',2);grid on;axis('square')
88 xlabel('t','FontSize',12);ylabel('u','FontSize',12);
89
90 print -dpng -r300 -f1 BVP_zermelo3.png;
91 print -dpng -r300 -f2 BVP_zermelo4.png;
92
```

- **Goal:** (Bryson page 66) determine the maximum radius orbit transfer in a given time t_f assuming a constant thrust rocket (thrust T).¹⁵
 - Must find the thrust direction angle $\phi(t)$
 - Assume a circular orbit for the initial and final times
- **Nomenclature:**
 - r – radial distance from attracting center, with gravitational constant μ
 - v, u tangential, radial components of the velocity
 - m mass of s/c, and \dot{m} is the fuel consumption rate (constant)
- **Problem:** find $\phi(t)$ to maximize $r(t_f)$ subject to:

$$\begin{aligned} \text{Dynamics : } \quad \dot{r} &= u \\ \dot{u} &= \frac{v^2}{r} - \frac{\mu}{r^2} + \frac{T \sin \phi}{m_0 - |\dot{m}|t} \\ \dot{v} &= -\frac{uv}{r} + \frac{T \cos \phi}{m_0 - |\dot{m}|t} \end{aligned}$$

with initial conditions

$$r(0) = r_0 \quad u(0) = 0 \quad v(0) = \sqrt{\frac{\mu}{r_0}}$$

and terminal conditions

$$u(t_f) = 0 \quad v(t_f) - \sqrt{\frac{\mu}{r(t_f)}} = 0$$

- With $\mathbf{p}^T = [p_1 \ p_2 \ p_3]$ this gives the Hamiltonian (since $g = 0$)

$$H = \mathbf{p}^T \begin{bmatrix} u \\ \frac{v^2}{r} - \frac{\mu}{r^2} + \frac{T \sin \phi}{m_0 - |\dot{m}|t} \\ -\frac{uv}{r} + \frac{T \cos \phi}{m_0 - |\dot{m}|t} \end{bmatrix}$$

¹⁵Thanks to Geoff Huntington

– Then $H_{\mathbf{u}} = 0$ with $\mathbf{u}(t) = \phi(t)$ gives

$$p_2 \left(\frac{T \cos \phi}{m_0 - |\dot{m}|t} \right) + p_3 \left(\frac{-T \sin \phi}{m_0 - |\dot{m}|t} \right) = 0$$

which gives that

$$\tan \phi = \frac{p_2(t)}{p_3(t)}$$

that can be solved for the control input given the costates.

- Note that this is a problem of the form on 6-6, with

$$\mathbf{m} = \begin{bmatrix} u(t_f) \\ v(t_f) - \sqrt{\frac{\mu}{r(t_f)}} \end{bmatrix} = 0$$

which gives

$$w = -r + \nu_1 u(t_f) + \nu_2 \left(v(t_f) - \sqrt{\frac{\mu}{r(t_f)}} \right)$$

- Since the first state r is not specified at the final time, must have that

$$p_1(t_f) = \frac{\partial w}{\partial r}(t_f) = -1 + \frac{\nu_2}{2} \sqrt{\frac{\mu}{r(t_f)^3}}$$

– And note that

$$p_3(t_f) = \frac{\partial w}{\partial v}(t_f) = \nu_2$$

which gives ν_2 in terms of the costate.

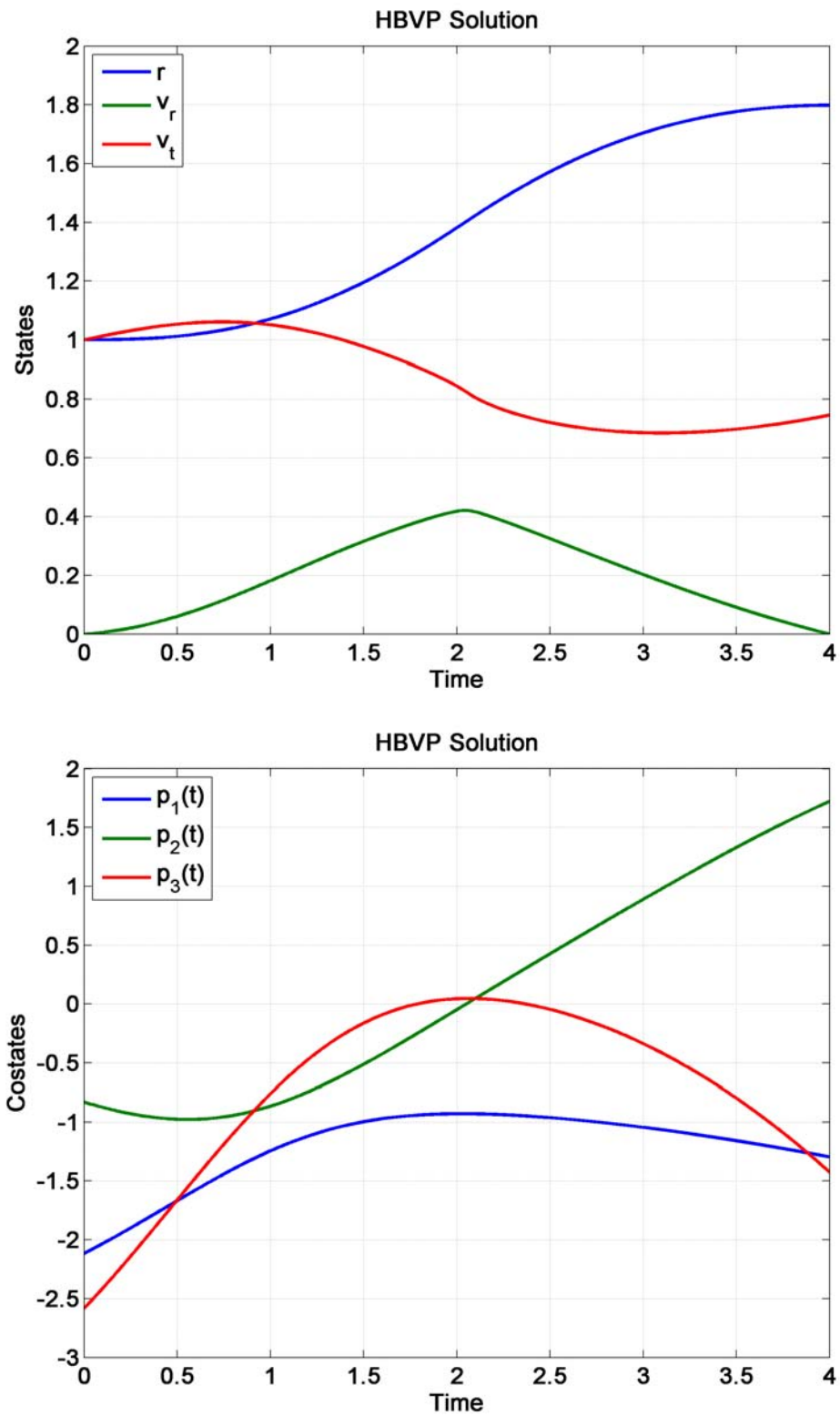


Figure 7.6: Orbit raising examples

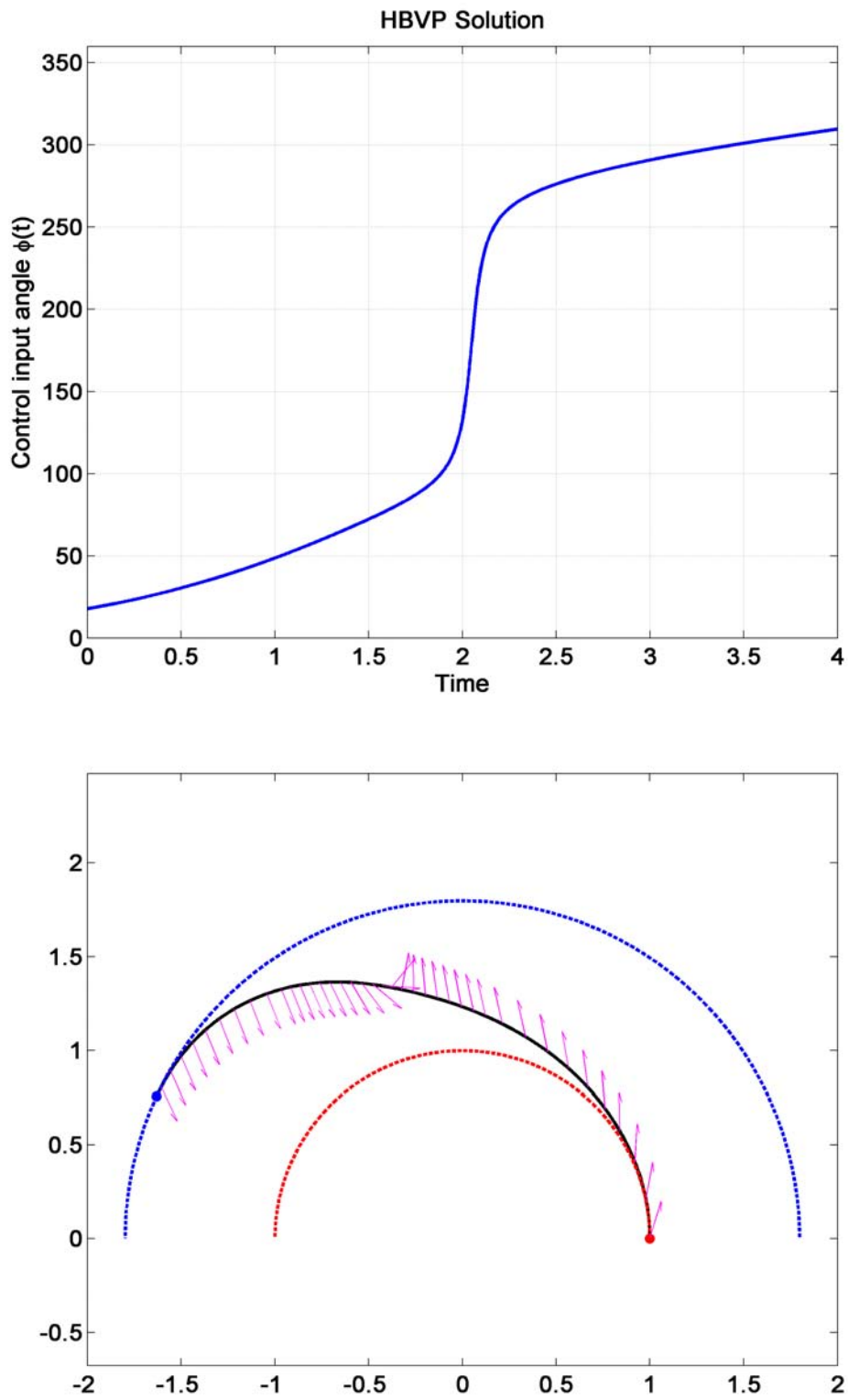


Figure 7.7: Orbit raising examples

Orbit Raising

```

1 %orbit_bvp_how created by Geoff Huntington 2/21/07
2 %Solves the Hamiltonian Boundary Value Problem for the orbit-raising optimal
3 %control problem (p.66 Bryson & Ho). Computes the solution using BVP4C
4 %Invokes subroutines orbit_ivp and orbit_bound
5 clear all;%close all;
6 set(0, 'DefaultAxesFontSize', 14, 'DefaultAxesFontWeight','demi')
7 set(0, 'DefaultTextFontSize', 14, 'DefaultTextFontWeight','demi')
8
9 %Fixed final time %Tf = 3.3155;
10 Tf = 4;
11 four=0; % not four means use bvp6c
12
13 %Constants
14 global mu m0 m1 T
15 mu=1; m0=1; m1=-0.07485; T= 0.1405;
16 %mu=1; m0=1; m1=-.2; T= 0.1405;
17
18 %Create initial Guess
19 n=100;
20 y = [ones(1,n); %r
21      zeros(1,n); %vr
22      ones(1,n); %vt
23      -ones(1,n); %lambda_r
24      -ones(1,n); %lambda_vr
25      -ones(1,n)]; %lambda_vt
26 x = linspace(0,Tf,n); %time
27 solinit.x = x;solinit.y = y;
28 %Set optimizer options
29 tol = 1E-10;
30 options = bvpset('RelTol',tol,'AbsTol',[tol tol tol tol tol tol],'Nmax', 2000);
31
32 %Solve
33 if four
34     sol = bvp4c(@orbit_ivp,@orbit_bound,solinit,options);
35     Nstep=40;
36 else
37     sol = bvp6c(@orbit_ivp,@orbit_bound,solinit,options);
38     Nstep=30;
39 end
40
41 %Plot results
42 figure(1);clf
43 plot(sol.x,sol.y(1:3,:), 'LineWidth',2)
44 legend('r','v_r','v_t','Location','NorthWest')
45 grid on;
46 axis([0 4 0 2])
47 title('HBVP Solution')
48 xlabel('Time');ylabel('States')
49
50 figure(2);clf
51 plot(sol.x,sol.y(4:6,:), 'LineWidth',2)
52 legend('p_1(t)','p_2(t)','p_3(t)','Location','NorthWest')
53 grid on;
54 axis([0 4 -3 2])
55 title('HBVP Solution')
56 xlabel('Time');ylabel('Costates')
57
58 ang2=atan2(sol.y([5],:),sol.y([6],:))+pi;
59 figure(3);clf
60 plot(sol.x,180/pi*ang2, 'LineWidth',2)
61 grid on;
62 axis([0 4 0 360])
63 title('HBVP Solution')
64 xlabel('Time');ylabel('Control input angle \phi(t)')
65 norm([tan(ang2)-(sol.y([5],:)/sol.y([6],:))'])
66
67 print -f1 -dpng -r300 orbit1.png
68 print -f2 -dpng -r300 orbit2.png
69 print -f3 -dpng -r300 orbit3.png
70
71 % Code below adapted inpart from Bryson "Dynamic Optimization"

```



```

72
73 dt=diff(sol.x);
74 dth=(sol.y(3,1:end-1)./sol.y(1,1:end-1)).*dt; % \dot \theta = v_t/r
75 th=0+cumsum(dth');
76 pathloc=[sol.y(1,1:end-1)'.*cos(th) sol.y(1,1:end-1)'.*sin(th)];
77
78 figure(4);clf
79 plot(pathloc(:,1),pathloc(:,2),'k-', 'LineWidth',2)
80 hold on
81 zz=exp(sqrt(-1)*[0:.01:pi]');
82 r0=sol.y(1,1);rf=sol.y(1,end);
83 plot(r0*real(zz),r0*imag(zz),'r--', 'LineWidth',2)
84 plot(rf*real(zz),rf*imag(zz),'b--', 'LineWidth',2)
85 plot(r0,0,'ro', 'MarkerFace','r')
86 plot(rf*cos(th(end)),rf*sin(th(end)),'bo', 'MarkerFace','b')
87 fact=0.2;ep=ones(size(th,1),1)*pi/2+th-ang2(1:end-1)';
88 xt=pathloc(:,1)+fact*cos(ep); yt=pathloc(:,2)+fact*sin(ep);
89 for i=1:Nstep:size(th,1),
90     pltarrow([pathloc(i,1);xt(i)], [pathloc(i,2);yt(i)], .05, 'm', '-');
91 end;
92 %axis([-1.6 1.6 -.1 1.8]);
93 axis([-2 2 -.1 1.8]);
94 axis('equal')
95 hold off
96
97 print -f4 -dpng -r300 orbit4.png;

```

```

1 function [dx] = orbit_ivp(t,x)
2 global mu m0 m1 T
3
4 %State
5 r = x(1);u = x(2);v = x(3);
6 lamr = x(4);lamu = x(5);lamv = x(6);
7
8 %Substitution for control
9 sinphi = -lamu./sqrt(lamu.^2+lamv.^2);
10 cosphi = -lamv./sqrt(lamu.^2+lamv.^2);
11
12 %Dynamic Equations
13 dr = u;
14 du = v^2/r - mu/r^2 + T*sinphi/(m0 + m1*t);
15 dv = -u*v/r + T*cosphi/(m0 + m1*t);
16
17 dlamr = -lamu*(-v^2/r^2 + 2*mu/r^3) - lamv*(u*v/r^2);
18 dlamu = -lamr + lamv*v/r;
19 dlamv = -lamu*2*v/r + lamv*u/r;
20
21 dx = [dr; du; dv; dlamr; dlamu; dlamv];

```

```

1 function [res] = orbit_bound(x,x2)
2 global mu m0 m1 T
3
4 %Initial State
5 r = x(1);u = x(2);v = x(3);
6 lamr = x(4);lamu = x(5);lamv = x(6);
7
8 %Final State
9 r2 = x2(1);u2 = x2(2);v2 = x2(3);
10 lamr2 = x2(4);lamu2 = x2(5);lamv2 = x2(6);
11
12 %Boundary Constraints
13 b1 = r - 1;
14 b2 = u;
15 b3 = v - sqrt(mu/r);
16 b4 = u2;
17 b5 = v2 - sqrt(mu/r2);
18 b6 = lamr2 + 1 - lamv2*sqrt(mu)/r2^(3/2);
19
20 %Residual
21 res = [b1;b2;b3;b4;b5;b6];

```
