

# VFI instability

```
In[54]:= x = 0; Remove["Global`*"]; DateList[Date[]] // Most  
Out[54]= {2020, 4, 27, 17, 8}
```

## Discrete-Time Growth - Log utility function

We analyze the discrete-time growth model with Cobb-Douglas production function and quadratic utility.

We choose parameters so that the steady state is  $k=1$ .

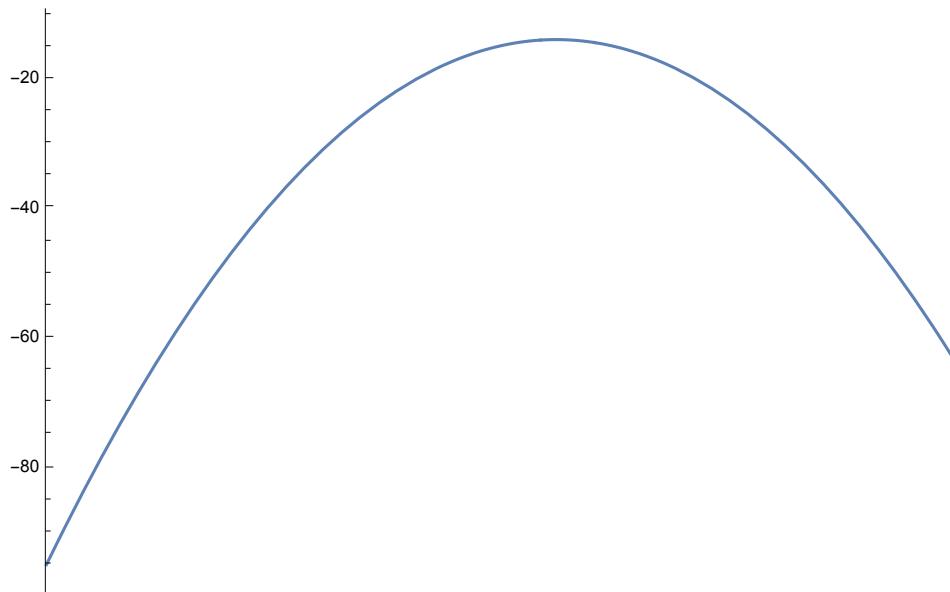
```
In[55]:= β = .95; α = .25; f[x_] = x + A x^α;  
A = (1-β) / (α β); kss = 1.; css = f[1]-1;
```

We choose the log utility function

```
util[x_] = Log[10., x];  
negc = Series[util[x], {x, .01, 2}] // Normal;  
u[x_] = If[x > .01, util[x], Evaluate[negc]]
```

Choose initial guess

```
In[61]:= valinit[x_] = -14 - 100 (x - 1)^2;  
Plot[valinit[x], {x, 0.1`, 1.7`}]
```



---

## Unstable example

```
In[63]:= xmin = 0.1; xmax = 1.5;
```

Specify nodes

```
In[64]:= nodes = Table[x, {x, .2, 1.5, .1}];  
Length[nodes]
```

```
Out[65]= 14
```

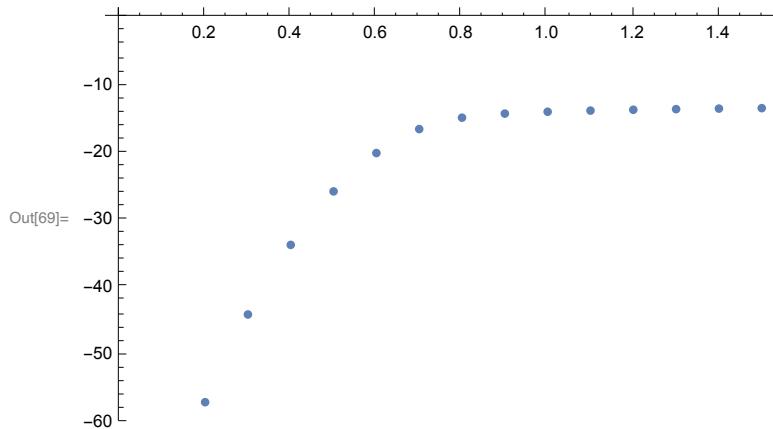
Define newval[x] which computes the new value of  $V[x]$  given by the RHS of the Bellman equation.

```
In[66]:= newval[x_] := FindMaximum[  
    u[c] + β val[f[x] - c], (* Objective *)  
    {c, css}, (* Initial guess *)  
    AccuracyGoal → 6][[1]]
```

Set  $\text{val}[x]$  to the initial guess and do first VFI

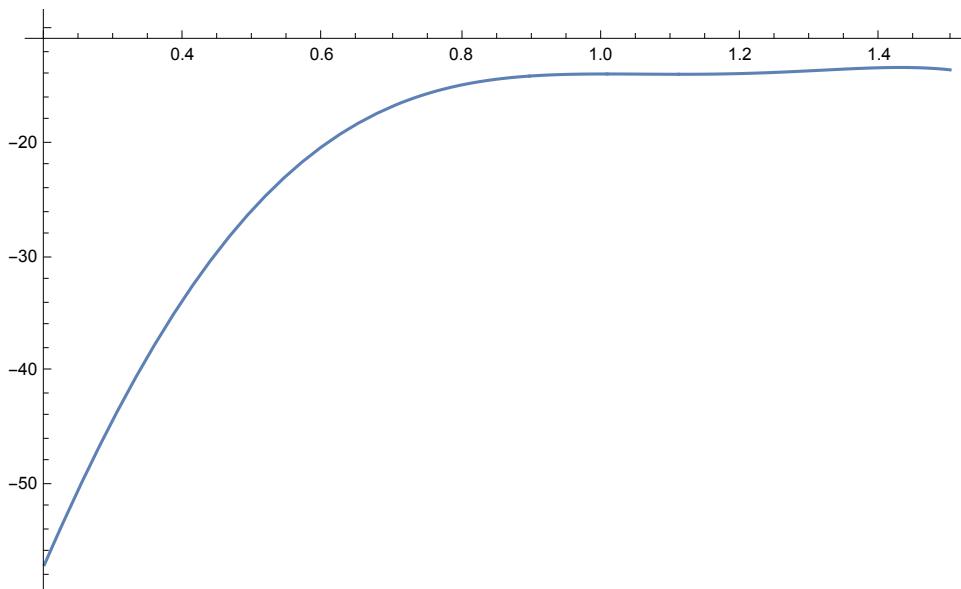
```
In[67]:= val[x_] = valinit[x];
```

```
In[68]:= newvs = Table[{nodes[[i]], newval[nodes[[i]]]}, {i, 1, Length[nodes]}];  
ListPlot[newvs]
```



Compute new value function

```
In[70]:= powers = Table[x^i, {i, 0, 5}];  
Clear[val]; val[x_] = Fit[newvs, powers, x]  
Plot[val[x], {x, 0.2` , 1.5`}, PlotRange -> All]  
Out[71]= -88.6899 + 168.997 x - 17.2121 x2 - 223.862 x3 + 197.348 x4 - 50.453 x5
```



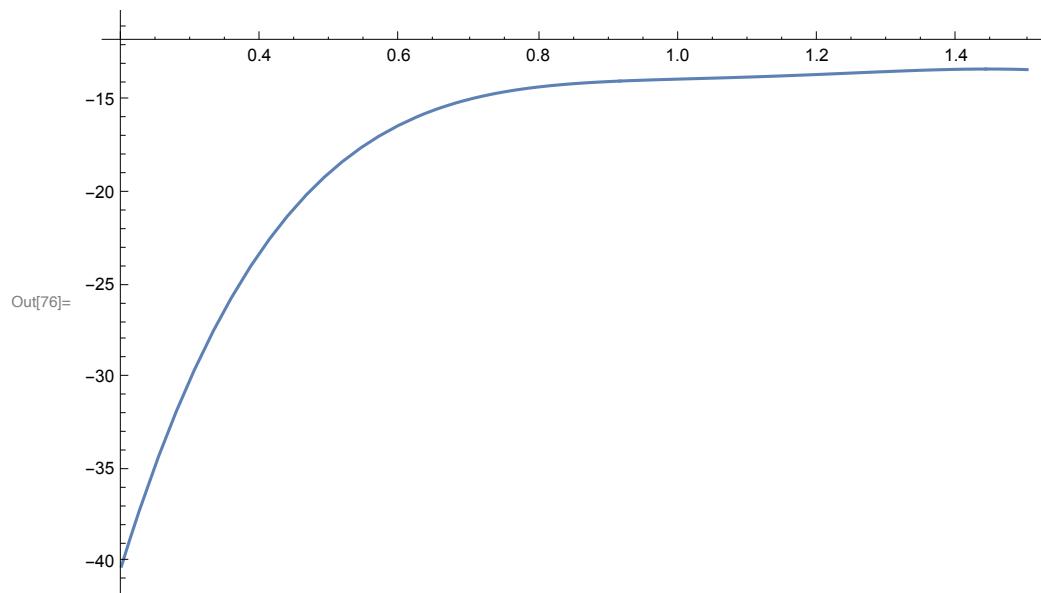
Define a value function iteration command

```
In[73]:= vfi :=  
(newvs = Table[{nodes[[i]], newval[nodes[[i]]]}, {i, 1, Length[nodes]}];  
Clear[val]; val[x_] = Fit[newvs, powers, x];  
Plot[val[x], {x, 0.2` , 1.5`}, PlotRange -> All])  
  
In[74]:= iter = 1;
```

```
In[75]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

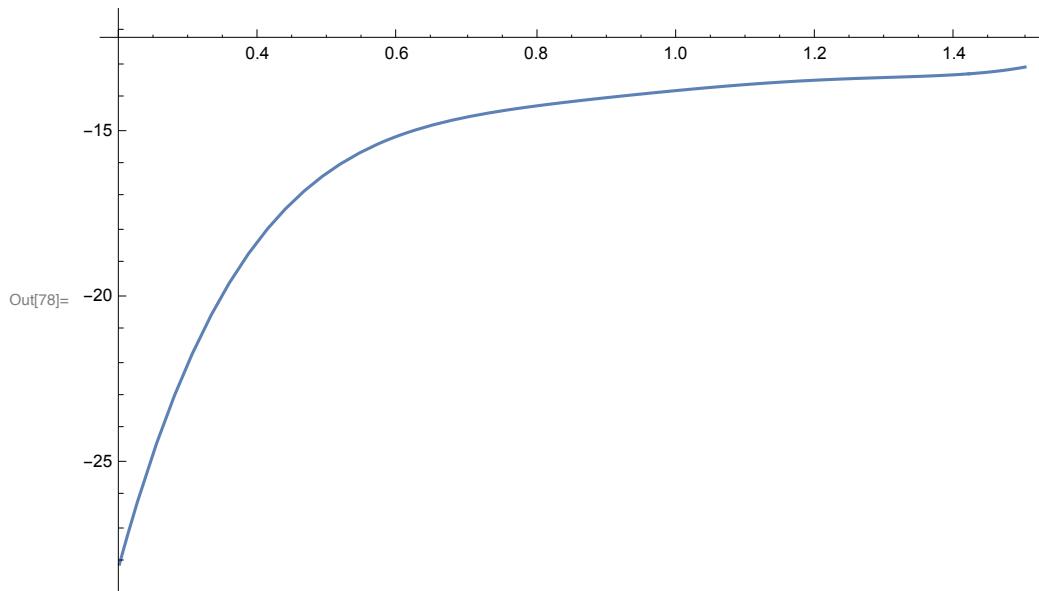
```
Out[75]= 2
```



```
In[77]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

```
iteration number:
```

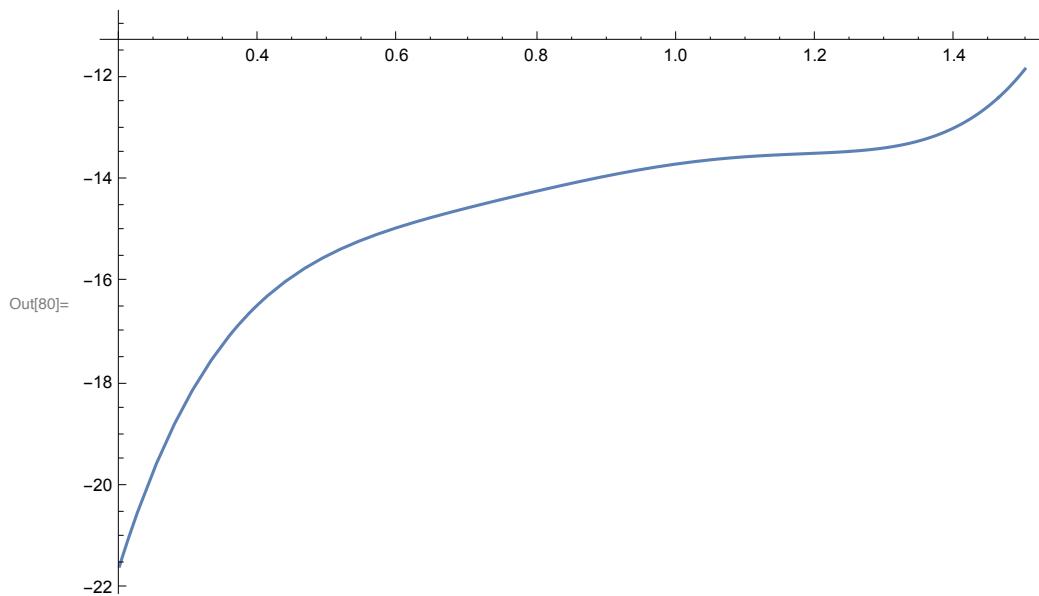
```
Out[77]= 3
```



```
In[79]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

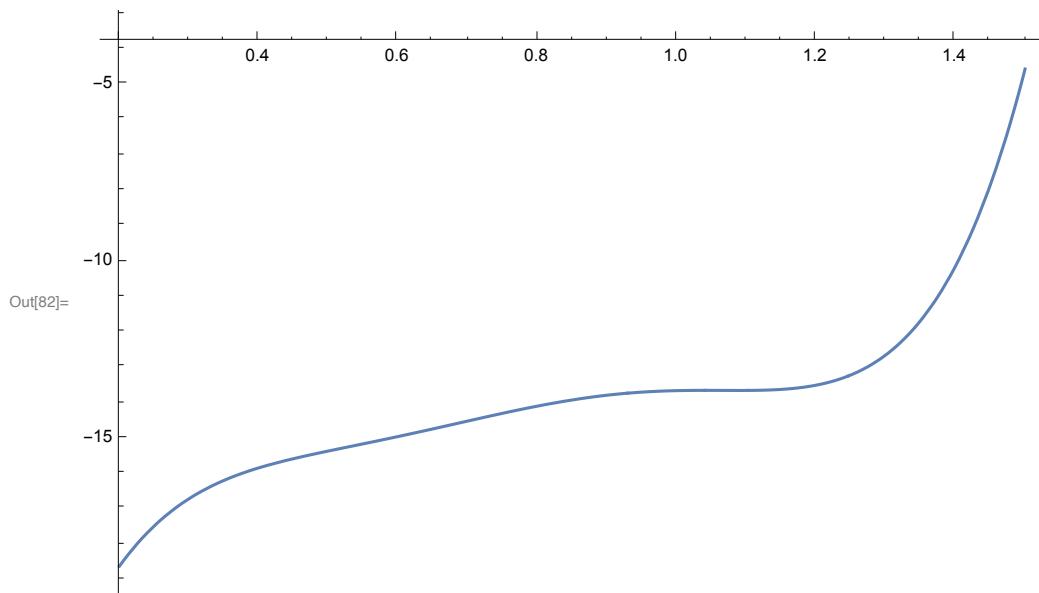
```
Out[79]= 4
```



```
In[81]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

```
iteration number:
```

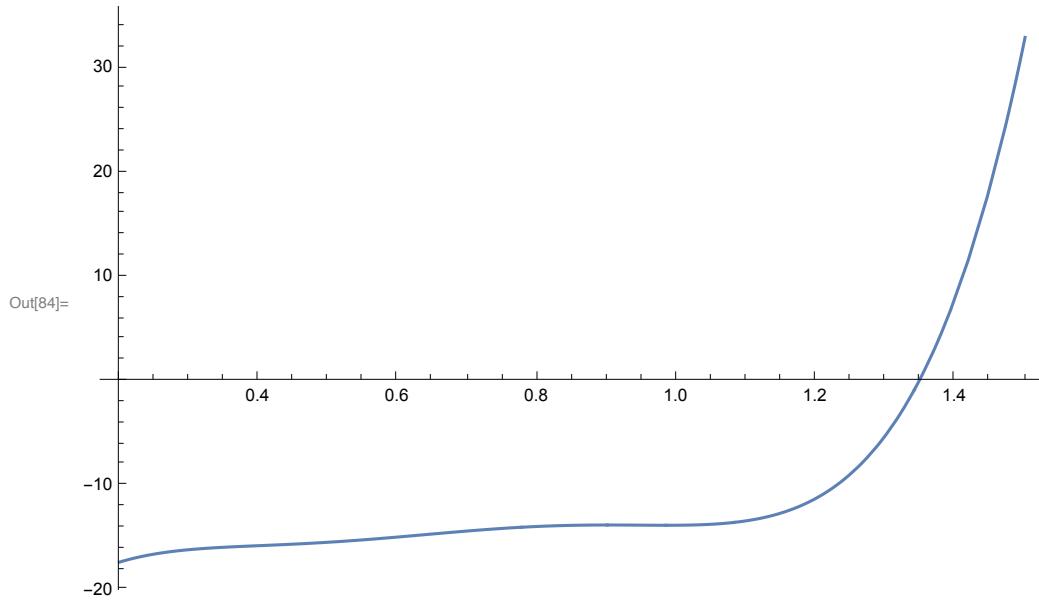
```
Out[81]= 5
```



```
In[83]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

```
iteration number:
```

```
Out[83]= 6
```



There is no next value function. The reason is that the previous iterate was sufficiently convex that some Bellman optimization problems were unbounded.

---

## Stabilize with constraint

Specify nodes

```
In[85]:= nodes = Table[x, {x, .2, 1.5, .1}];  
Length[nodes]  
Out[86]= 14
```

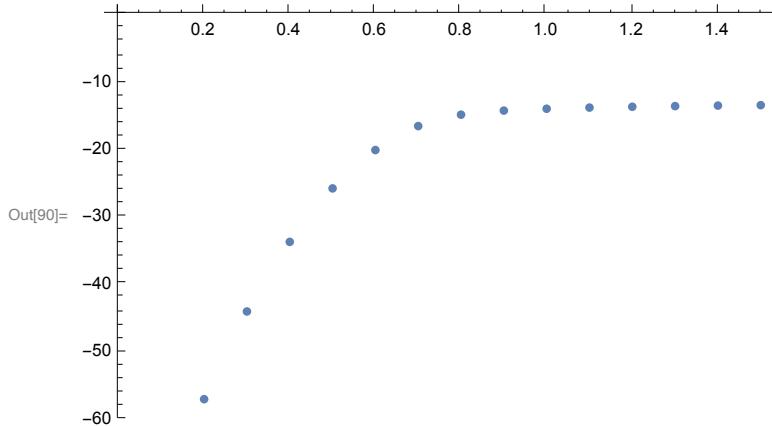
Define newval[x] which computes the new value of of  $V[x]$  given by the RHS of the Bellman equation.

```
In[87]:= newval[x_] := FindMaximum[  
{u[c] + β val[f[x] - c], xmin ≤ f[x] - c ≤ xmax}, (* Objective and constraint *)  
{c, css}, (* Initial guess *)  
AccuracyGoal → 6][[1]]
```

Set  $\text{val}[x]$  to the initial guess and do first VFI

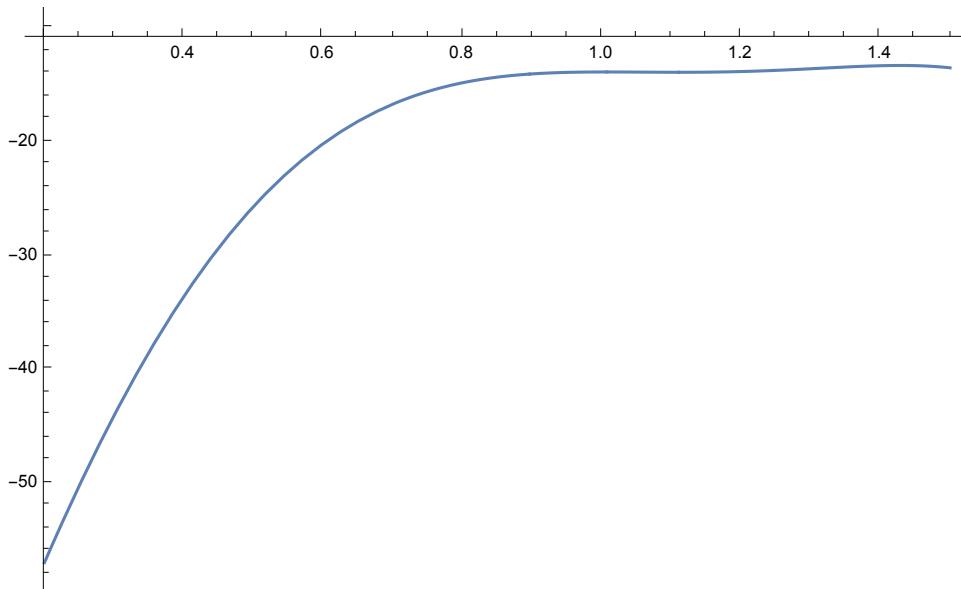
```
In[88]:= val[x_] = valinit[x];
```

```
In[89]:= newvs = Table[{nodes[[i]], newval[nodes[[i]]]}, {i, 1, Length[nodes]}];  
ListPlot[newvs]
```



Compute new value function

```
In[91]:= powers = Table[x^i, {i, 0, 5}];  
Clear[val]; val[x_] = Fit[newvs, powers, x]  
Plot[val[x], {x, 0.2` , 1.5`}, PlotRange -> All]  
Out[92]= -88.6899 + 168.997 x - 17.2121 x2 - 223.862 x3 + 197.348 x4 - 50.453 x5
```



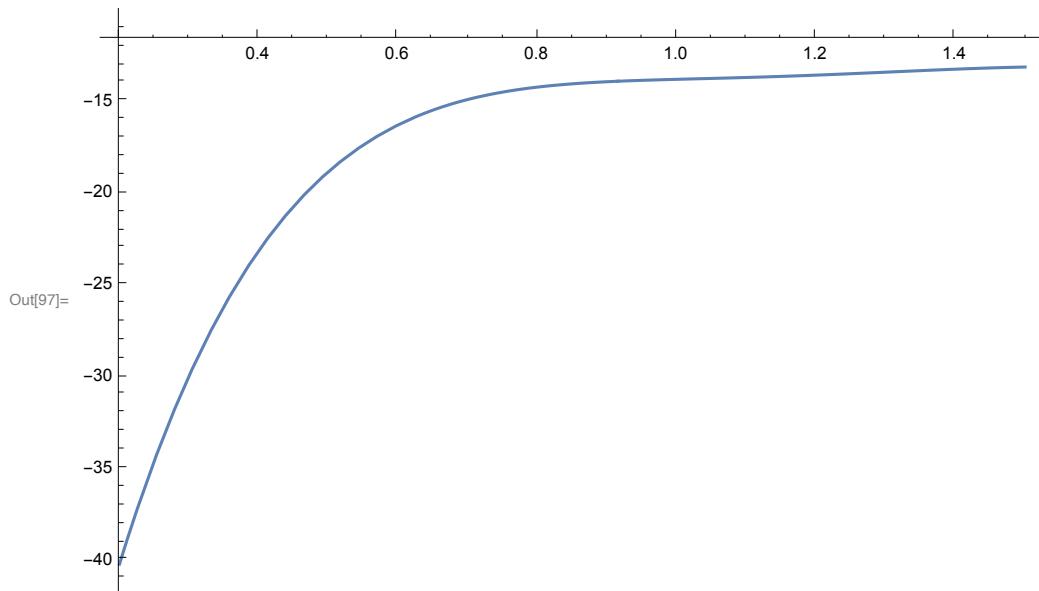
Define a value function iteration command

```
In[94]:= vfi :=
  (newvs = Table[{nodes[[i]], newval[nodes[[i]])}, {i, 1, Length[nodes]});
  Clear[val]; val[x_] = Fit[newvs, powers, x];
  Plot[val[x], {x, 0.2` , 1.5`}, PlotRange → All])
In[95]:= iter = 1;
```

```
In[96]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

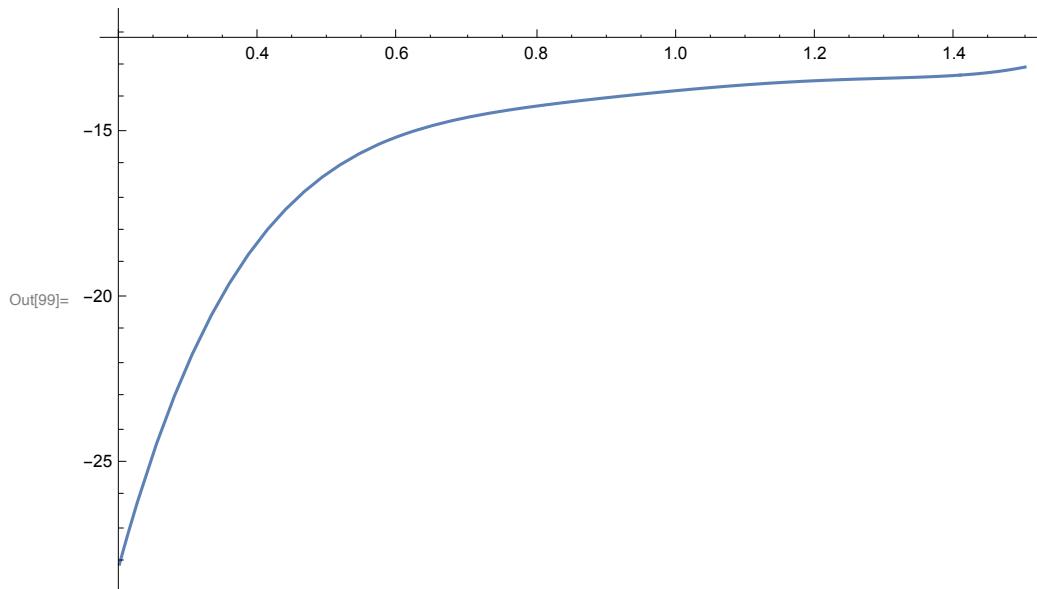
```
Out[96]= 2
```



```
In[98]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

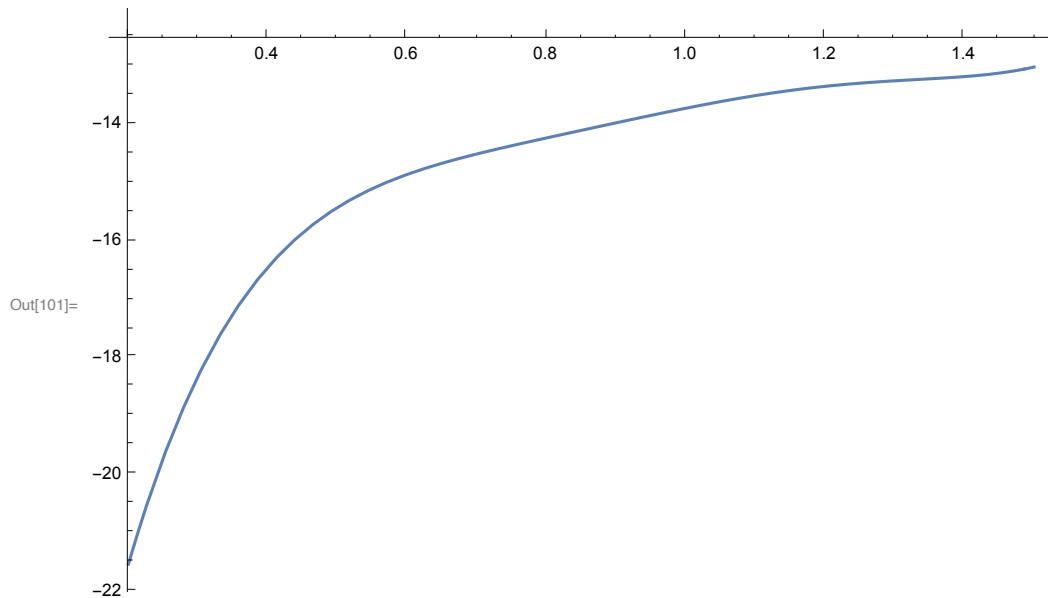
```
Out[98]= 3
```



```
In[100]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

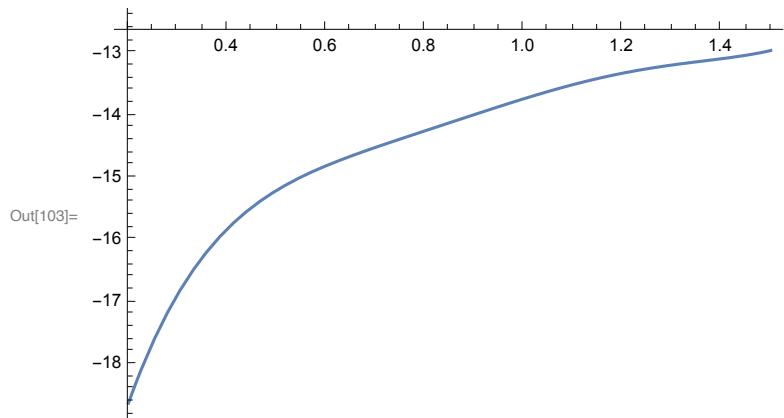
```
Out[100]= 4
```



```
In[102]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

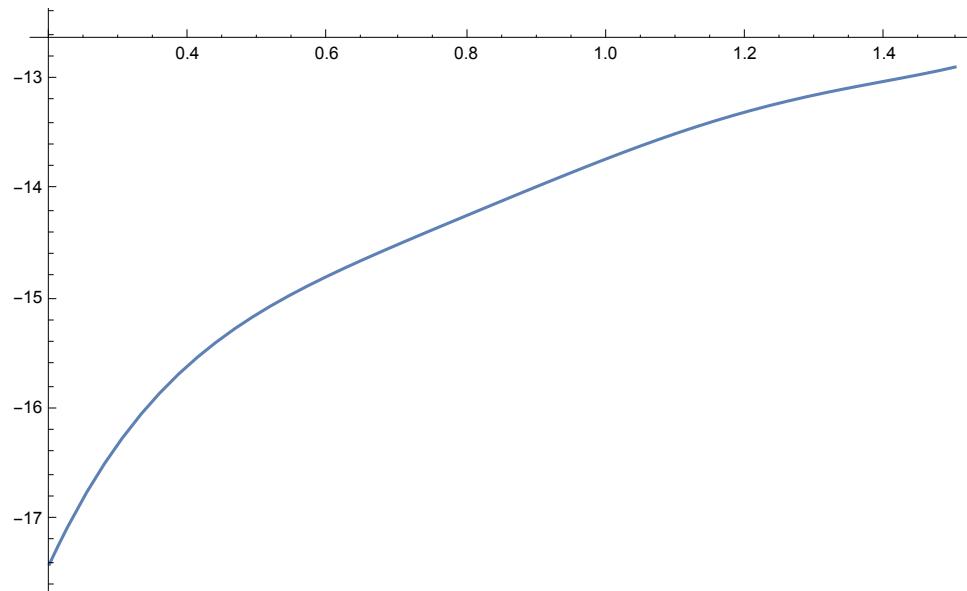
```
Out[102]= 5
```



```
In[104]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

```
iteration number:
```

```
Out[104]= 6
```

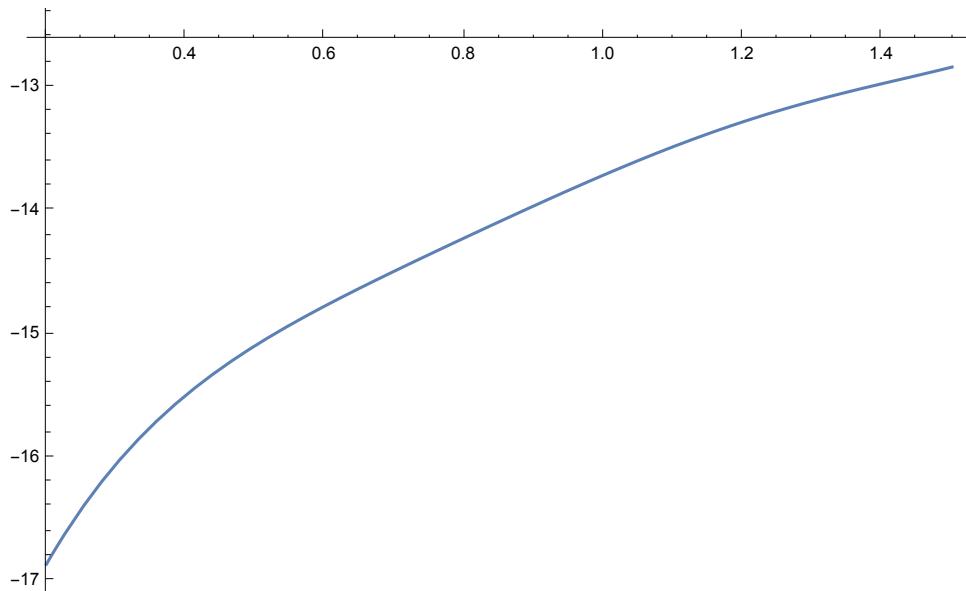


```
Out[105]= -15
```

```
In[106]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

```
iteration number:
```

```
Out[106]= 7
```

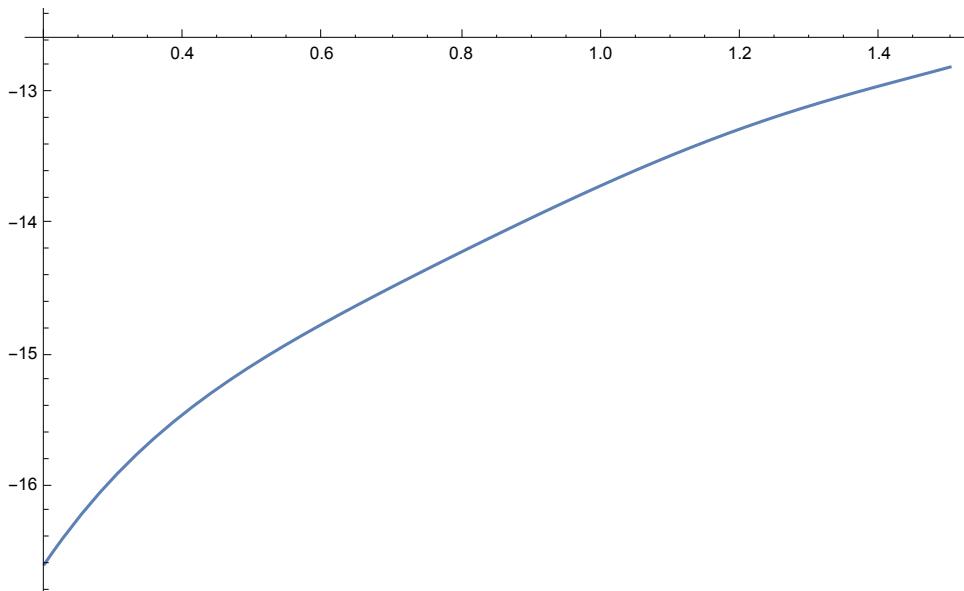


```
Out[107]=
```

```
In[108]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

```
Out[108]= 8
```



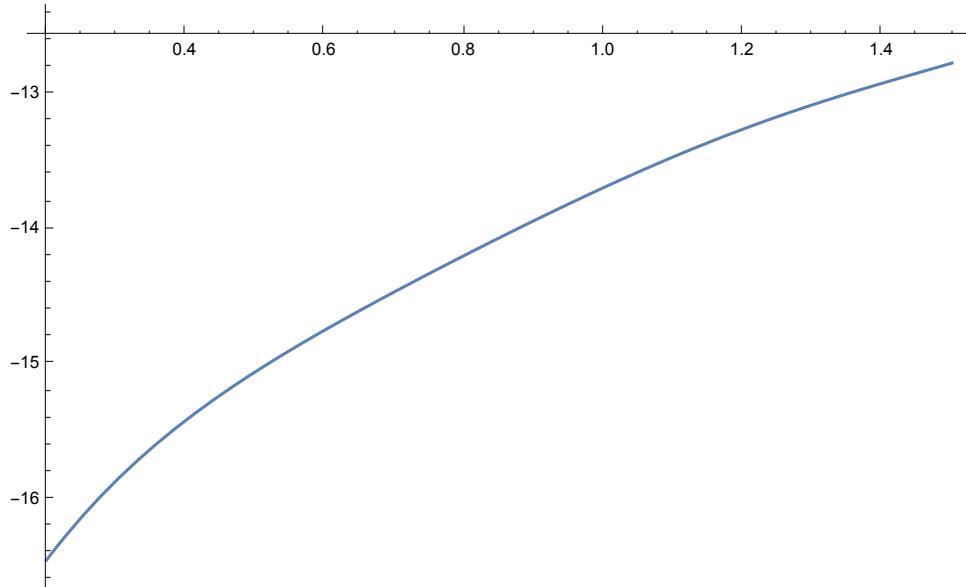
```
Out[109]=
```

```
In[110]:= Print["iteration number:"]; iter = iter + 1
```

```
vfi
```

```
iteration number:
```

```
Out[110]= 9
```

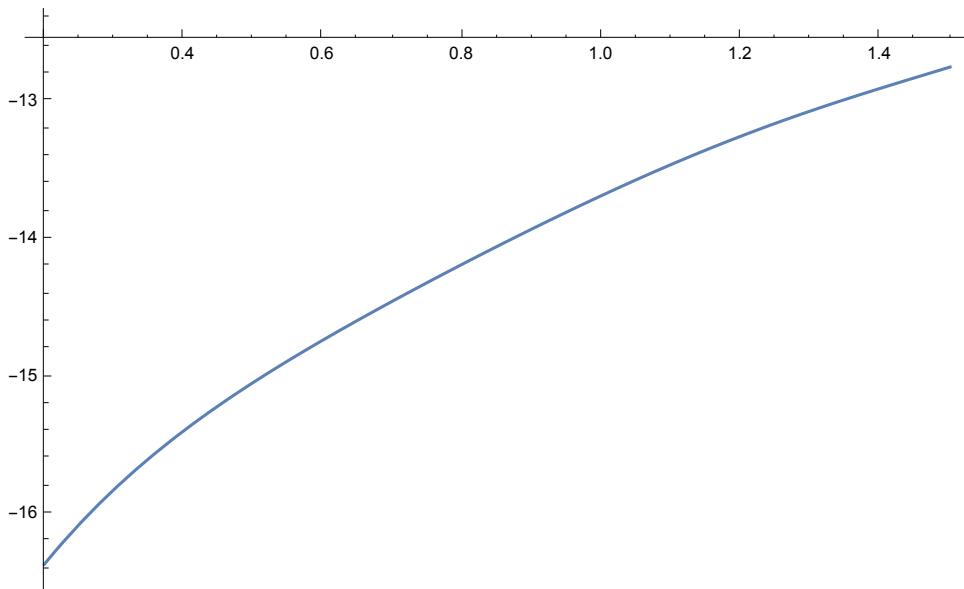


```
Out[111]=
```

```
In[112]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

```
iteration number:
```

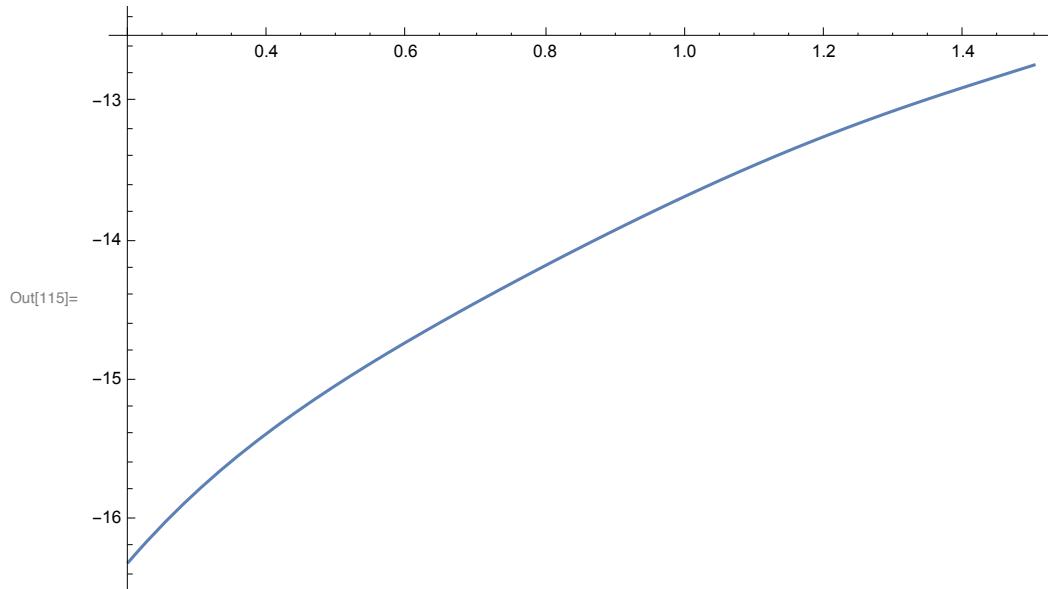
```
Out[112]= 10
```



```
In[114]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

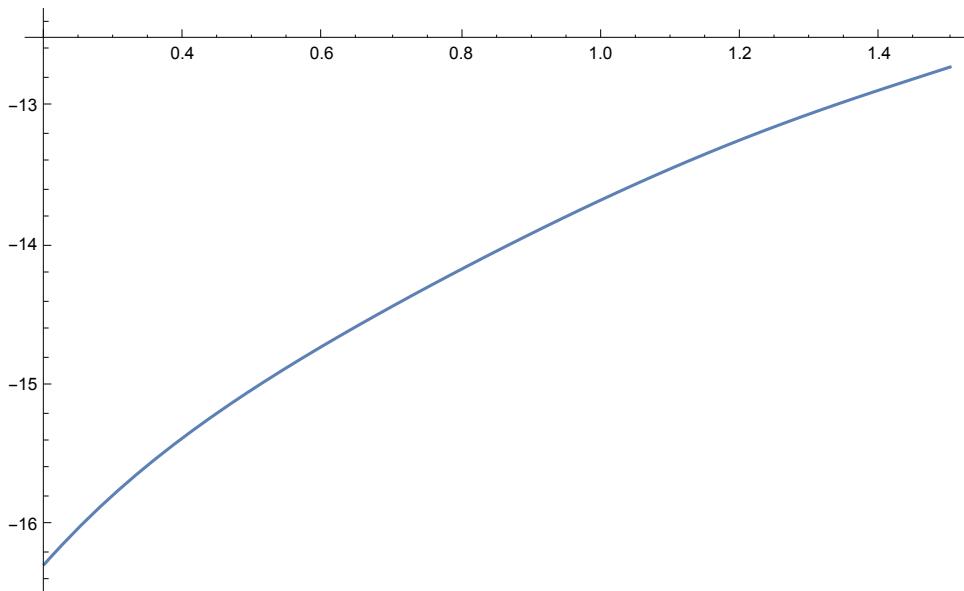
```
Out[114]= 11
```



```
In[116]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

```
iteration number:
```

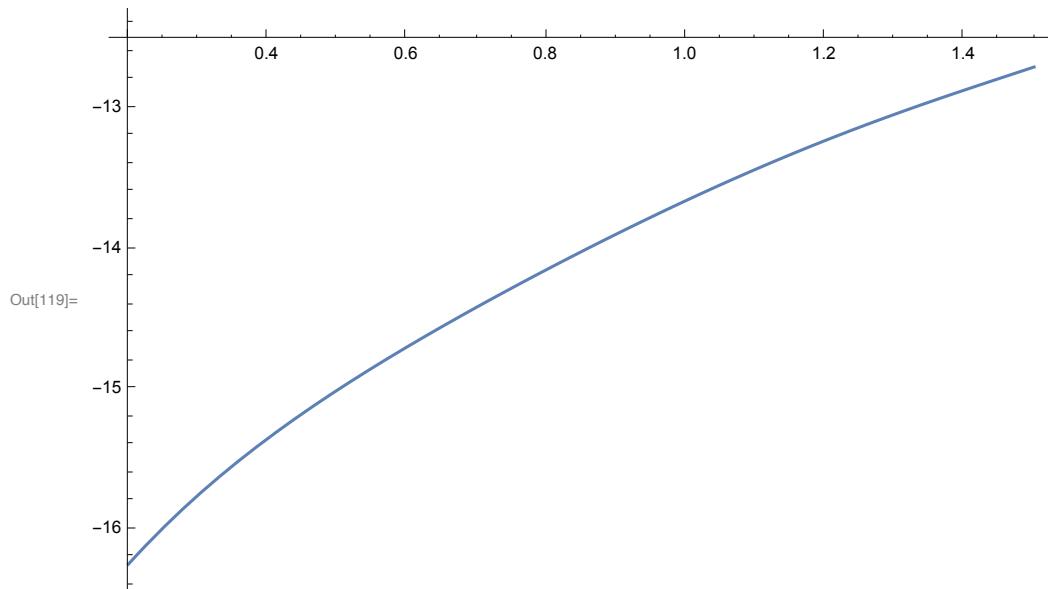
```
Out[116]= 12
```



```
In[118]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

iteration number:

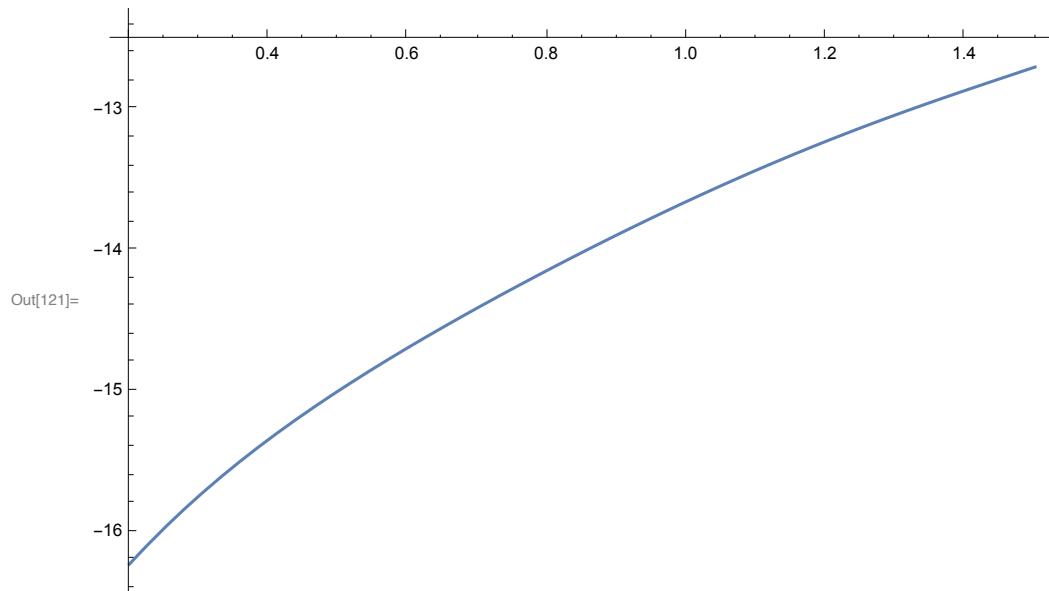
```
Out[118]= 13
```



```
In[120]:= Print["iteration number:"]; iter = iter + 1
vfi

iteration number:

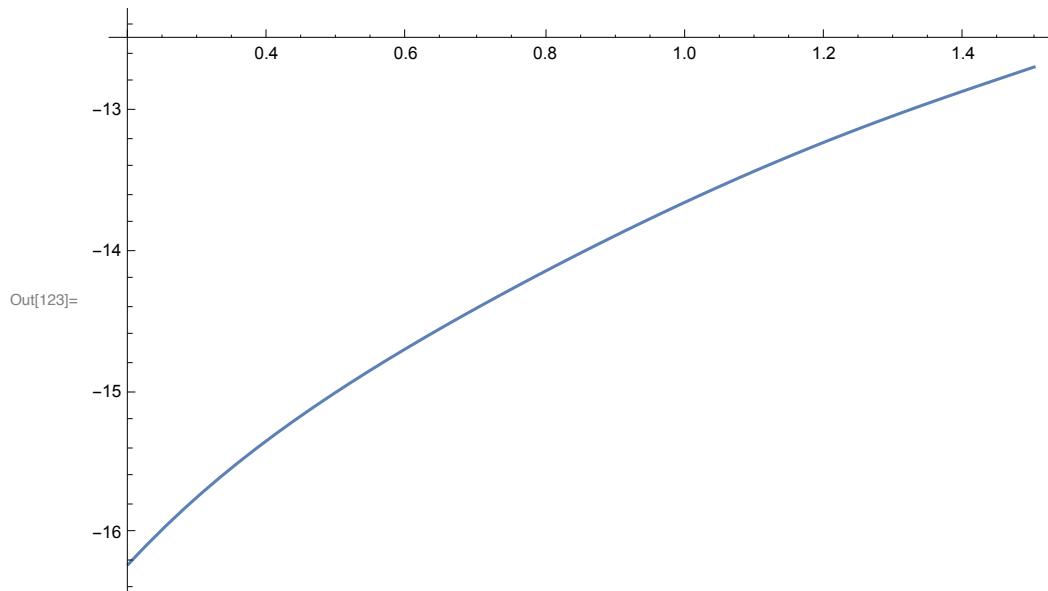
Out[120]= 14
```



```
In[122]:= Print["iteration number:"]; iter = iter + 1  
vfi
```

```
iteration number:
```

```
Out[122]= 15
```



## Lesson

Thou shalt not extrapolate.

Value function has no meaning outside the domain of definition

Extrapolations are often crazy

Add constraints forcing the state in the next period to be in the set of permissible states for the next period.