Higher Order Derivatives--Cobb Douglas

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We have the following general form for the Cobb-Douglas Production Function

$$Y(K, L) = K^{\alpha} \cdot L^{\beta}$$

The first order condition is

$$\frac{dY(K,L)}{dL} = (\beta) \cdot K^{\alpha} \cdot L^{\beta-1}$$

The derivative we have obtained is just another function. We can take additional derivatives with respect to this function.

$$\frac{\mathrm{d}^2 Y(K,L)}{dL^2} = (\beta) \cdot (\beta-1) \cdot K^\alpha \cdot L^{\beta-2}$$

Matlab symbolic toolbox gives us the same answer:

```
syms L K0 alpha beta f(L, K0, alpha) = K0^{(alpha)*L^{(beta)}}; first_deri = diff(f, L) first_deri(L, K0, alpha) = K_0^{\alpha} L^{\beta-1} \beta
```

```
second_deri = diff(diff(f, L),L)
```

```
second_deri(L, K0, alpha) = K_0^{\alpha} L^{\beta-2} \beta (\beta-1)
```

You can specify an additional parameter for the matlab *diff* function, if we want to take multiple derivatives:

```
syms L K0 alpha beta
f(L, K0, alpha) = K0^(alpha)*L^(beta);
% 5 for 5th derivative
tenth_deri = diff(f, L, 5)
```

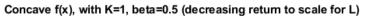
```
tenth_deri(L, K0, alpha) = K_0^{\alpha} L^{\beta-5} \beta (\beta-1) (\beta-2) (\beta-3) (\beta-4)
```

Curvature and Second Derivative, Concave Function

Let's graph out the second derivative when $\beta = 0.5$. The production function is concave (concave down). For a function that is twice continuously differentiable, the function is convex if and only if its second derivative is non-positive (never accelerating).

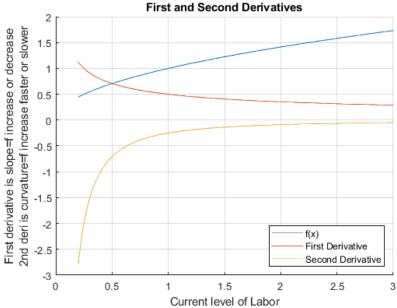
```
alpha = 0.5;
beta = 0.5;
K0 = 1;
% Note that we have 1 symbolic variable now, the others are numbers
syms L
f(L) = K0^(alpha)*L^(beta);
```

legend(['f(x)'], ['First Derivative'], ['Second Derivative'], 'Location','SE');



xlabel('Current level of Labor')

grid on



Curvature and Second Derivative, Convex Function

Let's graph out the second derivative when $\beta = 1.2$. The production function is convex (concave up). For a function that is twice continuously differentiable, the function is convex if and only if its second derivative is non-negative (never decelerating).

```
alpha = 0.5;
beta = 1.2;
K0 = 1;
% Note that we have 1 symbolic variable now, the others are numbers
f(L) = K0^{(alpha)*L^{(beta)}};
% Note here f diff1 L >= 0
f_diff1_L = diff(f, L)
f diff1 L(L) =
% Note here f diff2 L >= 0
f_diff2_L = diff(f, L, 2)
f_diff2_L(L) =
\frac{6}{25 L^{4/5}}
% Start figure
figure();
hold on;
% fplot plots a function with one symbolic variable
fplot(f, [0.1, 3])
fplot(f_diff1_L, [0.1, 3])
fplot(f_diff2_L, [0.1, 3])
title({'Convex f(x), with K=1, beta=1.2 (increasing return to scale for L)' 'First and Second I
ylabel({'First derivative is slope=f increase or decrease' '2nd deri is curvature=f increase fa
xlabel('Current level of Labor')
legend(['f(x)'], ['First Derivative'], ['Second Derivative'], 'Location','NW');
```

Convex f(x), with K=1, beta=1.2 (increasing return to scale for L)

grid on

