Module 9 Exercises | Joshua Shapiro | 19 June 2017

- 2. Why is this true? What is the minimum (unconstrained) value of f(x,y) = 3x + 4y?
- For any linear, unconstrained problem, the minimum will always be $-\infty$.
- 3. Go back to your calc book and find an example of a "hard to differentiate" function.
- Answer: $\frac{4x^2 + log(2^{tan(x^{sin(x)})})}{5x + cos(2^x)}$
- 4. What's an example of a function that's continuous but not differentiable? Consider the weird function f(x) where f(x) = 1 if x is rational and f(x) = 0 otherwise. Is this continuous? Differentiable?
- The absolute value function is not differentiable.
- The weird function is not continuous and therefore not differentiable.
- 5. Consider the function $f(x) = \frac{x}{\mu_1 \lambda x} + \frac{1 x}{\mu_2 \lambda(1 x)}$ Compute the derivative f'(x). Can you solve f'(x) = 0?

• Derivative below:
$$-f(x) = \frac{x}{\mu_1 - \lambda x} + \frac{1 - x}{\mu_2 - \lambda(1 - x)} \\ - \Rightarrow f'(x) = \frac{\mu_1 - \lambda * x + \lambda * x}{m u_1 - \lambda * x^2} + \frac{-\mu_2 + \lambda - \lambda * x - \lambda - \lambda * x}{(\mu_2 - \lambda + \lambda * x)^2} \\ - \Rightarrow f'(x) = \frac{(\mu_1)(\mu_2 - \lambda + \lambda * x)^2 - (\mu_2)(\mu_1 - \lambda * x)^2}{(\mu_1 - \lambda * x)^2(\mu_2 - \lambda + \lambda * x)^2}$$
• Solving for f'(x) = 0:
$$(\mu_1)(\mu_2 - \lambda + \lambda * x)^2 - (\mu_2)(\mu_1 - \lambda * x)^2 = 0$$

olving for
$$f'(\mathbf{x}) = \mathbf{0}$$
:
$$-\frac{(\mu_1)(\mu_2 - \lambda + \lambda * x)^2 - (\mu_2)(\mu_1 - \lambda * x)^2}{(\mu_1 - \lambda * x)^2(\mu_2 - \lambda + \lambda * x)^2} = 0$$

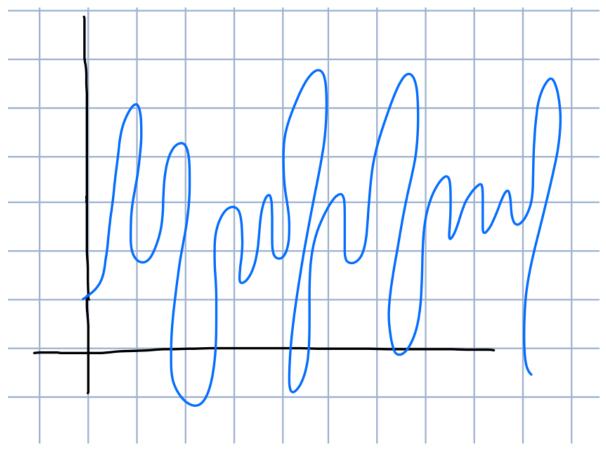
$$- \implies (\mu_1)(\mu_2 + \lambda(x - 1))^2 - (\mu_2)(\mu_1 - \lambda * x)^2 = 0$$

$$- \implies \mu_1 \lambda^2 - \mu_1^2 \mu_2 - 2\mu_1 \lambda \mu_2 + \mu_1 \mu_2^2 + x(4\mu_1 \lambda \mu_2 - 2\mu_1 \lambda^2) + x^2(\mu_1 \lambda^2 - \lambda^2 \mu_2) = 0$$

$$- \text{ ...skipping a few steps...}$$

$$- \implies x = \pm \sqrt{\frac{\mu_1 \mu_2 (\mu_1 - \lambda + \mu_2)^2}{\lambda^2 (\mu_1 - \mu_2)^2}} - \frac{4\mu_1 \lambda \mu_2 - 2\mu_1 \lambda^2}{2(\mu_1 \lambda^2 - \lambda^2 \mu_2)}$$

- 6. CODE: Download and execute BracketSearch.java.
 - What is the running time in terms of M and N?
 - If we keep MN constant (e.g., MN = 24), what values of M and N produce best results?
- The run time is O(MN)
- For the function in BracketSearch, M = 12 and N = 2 produced best results, with a minimum found at 2.5001 (actual minimum at 2.5).
- 7. Draw an example of a function for which bracket-search fails miserably, that is, the true minimum is much lower than what's found by bracket search even for large M and N.
- Any function with many peaks and valleys will cause this to fail.



- 8. What is the number of function evaluations in terms of M and N for the bracket-search algorithm?
- There are (M+1)N + 1 function evaluations.
- 9. What is the number of function evaluations in terms of M and N for the 2D bracket-search algorithm? How does this generalize to n dimensions?
- There are $((M+1)^2+1)N$ function evaluations for the 2D bracket-search algorithm. • There are $((M+1)^n+1)N$ function evaluations for the n-dimensional bracket-search algorithm.
- There are $((M+1)^n+1)N$ function evaluations for the n-dimensional bracket-search algorithm.
- 10. CODE: Add code to MultiBracketSearch.java to find the minimum of: $f(x_1, x_2) = (x_1 4.71)^2 + (x_2 3.2)^2 + 2(x_1 4.71)^2 * (x_2 3.2)^2$
 - With M = 6 and N = 4, x1 = 4.691358024691359, x2 = 3.2098765432098757, numFuncE-vals=120
- 11. CODE: Modify BracketSearch2.java to use the proportional-difference stopping condition.
- With $M=10,\,N=4,\,bestx=4.72,\,bestf=2.5001,\,prevBestf=2.50809999999999998$
- 16. CODE: Download and execute GradientDemo.java
 - How many iterations does it take to get close to the optimum?
 - What is the effect of using a small α (e.g., $\alpha = 0.001$)?
 - In the method nextStep(), print out the current value of x, and the value of xf'(x) before the update.
 - Set $\alpha = 1$ Explain what you observe.
 - What happens when $\alpha = 10$?
 - It takes around 150 iterations to get close to the optimum.

- A smaller alpha requires more iterations to get to the optimum.
- Alpha is too large when set to 1 and jumps over the minimum, never able to find it. Its jumps are always consistent, so it alternates between the same values.
- When alpha is set to 10, it also is never able to find the minimum. However, the jumps are not consistent and always moves farther away from the minimum.
- 17. CODE: Download GradientDemo2.java and examine the function being optimized.
 - Fill in the code for computing the derivative.
 - Try an initial value of x at 1.8. Does it converge?
 - Next, try an initial value of x at 5.8. What is the gradient at the point of convergence?
- It converges at the global minimum with a value of x = 1.8.
- When x = 5.8 it only finds a local minimum