

Stat 310, Part II, Optimization. Homework 2.

Problem 1: (computation; line search method, also version of Problem 3.1 in textbook)

Write a program that implements the backtracking (Armijo) line search with modified LDL' factorization (first, write a pseudocode of the entire method, this will help in your implementation. Report the pseudocode in your homework). Choose B_k to be the exact Hessian. Apply it to solve Fenton's function (from preceding homework) starting at both $[3 \ 2]$ and $[3 \ 4]$. Experiment with the parameters or by designing your own rules. Report the total number of linear systems solved and the total number of function evaluations.

Problem 2: (computation; "the power of sparse linear algebra").

Apply the algorithm above to solve the problem "cute" which is posted on my website. **Be sure the implementations supports sparse linear algebra.** Solve versions of the problem of increasing size, initialized at a vector of all ones, to a point which seems reasonable (on my laptop, evaluations of Hessian with INTVAL is very fast up to $n=10000$, on UNIX, $n=1000$ should be fast enough). Plot the number of function evaluations and number of linear systems solved as a function of the problem size.

Calling sequences for my function are (for example)

```
[f,g,H]=cute_wrap(ones(10000,1),2);
```

```
[f,g,H]=cute_wrap(x,2);
```

Problem 3: (divided differences, derivative calculations).

Consider the following divided difference approach (called "central differences") by which to approximate the derivative of the function $f(x): R \rightarrow R$, $f \in C^4$, (the function f is 4 times continuously differentiable):

$$\left. \frac{df}{dx} \right|_{x_0} \approx \frac{f(x_0 + h) - f(x_0 - h)}{2h}.$$

I. Use Taylor series expansions to justify that

$$\left| \left. \frac{df}{dx} \right|_{x_0} - \frac{f(x_0 + h) - f(x_0 - h)}{2h} \right| = O(h^2)$$

II. Using the same argument as in class, what is the optimal perturbation amount h with respect to the machine precision $\varepsilon \approx 1e-16$? (that is, the h that

produces the minimum value of the error). Is this minimum error smaller than the one for forward differences (the case covered in class?)

- III. Would you recommend central differences for approximating ∇f , $f : R^n \rightarrow R$ (by doing the perturbation above in each component) over the forward differences approach that we have discussed in class? Please explain your conclusion.