## Referee's report:

## Adaptive barrier strategies for nonlinear interior methods Jorge Nocedal, Andreas Wächter and Richard A. Waltz

The authors present strategies for adaptively adjusting the barrier parameter in interior-point methods for nonlinear programming, which in my opinion is a very interesting line of research. The investigation is done by an extension of predictor-corrector methods for linear programming. The reduction of  $\mu$ is measured by progress with respect to a partially linearized approximation of the residual of the nonlinear equations that firm the first-order optimality conditions. The authors embed their scheme in a "classical" reduction scheme to ensure convergence. They also include a discussion on the suitability of the corrector step for linear programming, when the initial point is not particularly well chosen.

The convergence strategy resembles spacer steps [Lue84], which is not a new idea. The idea is to use a known method to guarantee convergence (monotone mode) and make sure that the proposed method (free mode) does not inhibit convergence. A feature here is that the free steps do not contribute at all to the convergence. We could replace the l monotone steps by l steps of (almost) any kind. I do not find this very appealing. What is there to say that the approach that the authors suggest is particularly good? For example, would the minimizing  $\sigma$  in (4.2) be a particularly good choice?

I do not fully understand the convergence argument after the globalization framework. The authors claim that "... it is clear that  $\Phi_k \to 0$ ". As I state below in item 4, I find the use of k misleading. It is not totally clear what work is involved in each step. Is it true with no assumptions that all limit points satisfy the first-order optimality conditions, as stated in the following sentence? I would prefer a more specific globalization discussion.

## Specific comments

- 1. It is mentioned that [7,20] provide certain convergence results to stationary points. Isn't that exactly what is done here too? The merit function is the residual of the first-order optimality conditions.
- 2. The authors talk about solving (3.3). Yes, but what if no solution exist, or if the inertia is wrong? This is explained much later.

3. Page 5, line (3.5)+4.

I do not understand what "conversely" refers to. Is it  $\sigma$  small implies good progress?

- 4. I find the use of k confusing in the table of the globalization framework. I would think of k being one iteration, i.e., one set of (x, y, z). I assume that this is what is tabulated in the figures. However, for the monotone mode, you may perform many iterations before the convergence criteria is satisfied. Hence, the update of k by one appears very odd to me. (As it is now, I would solve the problem with  $k \leq 2$  if  $\kappa$  is sufficiently small.)
- 5. It appears to me that there is a minus sign missing in (7.1).
- 6. I do not follow the discussion in the paragraph following (7.3). It is the right-hand side of (3.3), and not (2.5), as far as I can see, since  $\mu = 0$  for the affine-scaling step.
- 7. It is not clear to me that the corrector step is applied at the primaldual point. Rather, the corrector step and the affine scaling step are both independent of  $\mu$ . The corrector step is aimed at improving the affine scaling step. To me, they form one entity aimed at optimality, and the step incurred by  $\mu$  is aimed at centrality. It is not clear to me that the corrector step of Figure 4 is unfavorable. If we add the affinescaling step and the corrector step, we are very close to the optimum. Ideally, they would give the optimum together, so I do not see why this step is unfavorable.
- 8. I do not understand the results of Table 1. For LP, we would expect a dual variable to diverge if the corresponding primal inequality constrained has the satisfied with equality on the feasible region, and symmetrically for the primal variables. I am not very familiar with the NETLIB test set, but this may be the case for FORPLAN. However, I would not expect the objective function values to diverge. I am not totally sure what disabling the initial point procedure in PCx means. The initial point of ones does not appear very bad to me, so I do not understand why PCx would fail for this choice of initial point.
- 9. If predictor-corrector methods for linear programs are so sensitive for the choice of starting point, is this the case also for nonlinear programs? I cannot see described how you go about defining initial strictly positive x and z. I assume that you may have to add slack variables,

or something similar, if the constraint is not strictly positive, and then the initial value might be crucial.

## References

[Lue84] D. G. Luenberger. Linear and Nonlinear Programming. Addison-Wesley Publishing Company, Reading, second edition, 1984. ISBN 0-201-15794-2.