PS 7

Due: Mon, Apr 13

1: By the Book Section 9.4, Problem 6. For part (b), it may help to refer to a homework problem earlier in the semester!

Answer: For the first part: Newton solves a linear approximation, and in this case the original equation is linear. Therefore, Newton converges in one step.

For the second part, consider convergence to the multiple root of $x^3 = 0$, which is linear. We did this example early in the semester.

2: Naive Newton Consider the nonlinear system of equations

$$x^2 + xy^2 = 9$$
$$3x^2y - y^3 = 4$$

Fill in the following Newton iteration code:

```
\label{eq:fork} \begin{split} & \textbf{for } k = 1{:}20 \\ & F = \% \ \textit{Your code here} \\ & \textbf{if norm}(F) < rtol \\ & \textbf{break}; \\ & \textbf{end} \\ & J = \% \ \textit{Your code here} \\ & dx = J \backslash F; \\ & x = x{-}dx; \\ & \textbf{end} \end{split}
```

Run your code with an initial guess of (1,1) and a residual norm tolerance of 10^{-12} . Do you see quadratic convergence?

Answer: The code follows:

```
\begin{array}{l} x = [1; \ 1]; \\ \textbf{for} \ k = 1:20 \\ F = [\ x(1)^2 + x(1)*x(2)^2 - 9; \\ 3*x(1)^2*x(2) - x(2)^3 - 4\ ]; \\ \textbf{fprintf}(`\%d:..\%e\n', \ k, \ \textbf{norm}(F)); \\ \textbf{if} \ \ \textbf{norm}(F) < 1e-12 \\ \textbf{break}; \\ \textbf{end} \end{array}
```

$$\begin{array}{ll} J = [\ 2*x(1) + x(2)^2, & 2*x(1)*x(2); \\ 6*x(1)*x(2), & 3*x(1)^2 - 3*x(2)^2]; \\ dx = J\backslash F; \\ x = x-dx; \\ \textbf{end} \end{array}$$

The residual was

1: 7.280110e+00 2: 4.875347e+01 3: 1.395474e+01 4: 2.916736e+00

5: 1.355139e-01 6: 6.031975e-05

7: 1.058961e-10

8: 5.024296e-15

From steps 5–8, convergence is clearly quadratic (with roundoff dominating by step 8); that is, $e_6 \approx e_5^2$ and $e_7 \approx e_6^2$.

3: Continue with Care Consider the boundary value problem

$$v''(x) + \gamma \exp(v(x)) = 0, \quad 0 < t < 1$$

 $v(0) = v(1) = 0$

discretized via finite differences on a mesh with 100 equally spaced points; see example 9.3 in the book.

- Write a code to find v for a range of γ values from 1 to 3.5 (use gammas = linspace(1,3.5) to generate the mesh). For the first value of γ , you should use an initial guess of v=0; for subsequent values, use the value of γ at the previous iterate. Plot all your solutions together on a single plot.
- For all γ in the given range, the Jacobian matrix at the solution remains negative definite. Plot $\lambda_{\max}(J(x^*))$ (the eigenvalue closest to zero) as a function of γ . What do you notice?
- Try running your code again, this time going up to a maximum value of 4 rather than 3.5. What happens?

Note: You may start from the following code

```
n = 100;

h = 1/(n+1);

T = diag(ones(n-1,1),-1) + diag(one(n-1,1),1) - 2*eye(n);

v = zeros(n,1);
```

If n was very large, we might want to use a sparse matrix¹, but it's probably not worth it in this case.

Answer: My code follows:

```
n = 100;
h = 1/(n+1);
T = diag(ones(n-1,1),-1) + diag(ones(n-1,1),1) - 2*eye(n);
v = zeros(n,1);
x = linspace(0,1,n+2);
x = x(2:end-1);
gammas = linspace(1,3.5);
lambdas = [];
figure(1);
hold on
for gamma = gammas
  for k = 1:10
    F = h^- - 2 * (T*v) + \mathbf{gamma} * \mathbf{exp}(v);
    J = h^- - 2 * T + \mathbf{diag}(\mathbf{gamma} * \mathbf{exp}(v));
    v = v - J \setminus F;
  end
  lmax = max(eig(J));
  lambdas = [lambdas, lmax];
  \mathbf{plot}(\mathbf{x}, \mathbf{v})
end
figure(2);
plot(lambdas)
```

It would probably be smarter to put a convergence criterion on the loop, but this is fast enough that I was willing to be lazy. The code blows up if

 $^{^{1}\}mathrm{I'd}$ probably switch to a more accurate discretization method, first, but that's a topic for CS 4210.

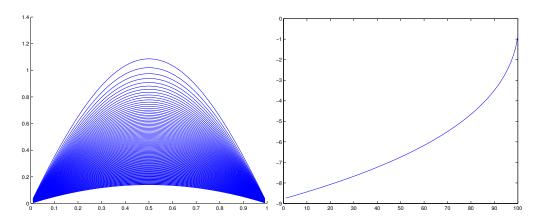


Figure 1: Solutions to $v'' + \gamma \exp(v) = 0$ for $\gamma \in [1, 3.5]$ (left) and maximum eigenvalue for the discrete Jacobian vs γ (right). For γ even a little larger than 3.5, the Jacobian matrix becomes singular at a turning point (and the solution cannot be continued to larger γ values).

the upper bound is much greater than 3.5. The plots show that the Jacobian likely becomes singular for γ a bit above 3.5; that γ represents a turning point for the solution manifold, and the solution branch does not exist for larger γ . This shows up when actually trying to run the computation because the Newton iteration starts to diverge, despite the continuation strategy.