Math 5329, Test I

Name Key

1. a. Find $T_n(x)$, the Taylor series of degree n for the function f(x) = ln(1+x), expanded around c=0.

(Hint: $f^{(n)}(x) = (-1)^{n-1}(n-1)!/(1+x)^n$, for $n \ge 1$.)

$$T_n(x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \frac{x^4}{6} + \dots + \frac{x^n}{n}$$

b. Find $E_n(x)$, the error in $T_n(x)$, and find a reasonable upper bound on $E_n(1)$.

$$2\frac{\left(E_{\Lambda}(x)\right) = \left(\frac{1}{(1+\epsilon)}\frac{1}{n+1}\right)}{\left(E_{\Lambda}(x)\right) = \left(\frac{1}{(1+\epsilon)}\frac{1}{n+1}\right) = \left(\frac{1}{(1+\epsilon)}\frac{1}{n+1}\right) = \frac{1}{(1+\epsilon)}\frac{1}{n+1}$$

c. Estimate the number of terms n required for $T_n(x)$ to approximate f(1) = ln(2) to 5 decimal places accuracy.

d. Would you expect roundoff error to be a serious concern in (c)? Why or why not?

(Hint: $1 + 1/2 + 1/3 + 1/4 + 1/5 + ... + 1/n \approx ln(n)$, for large n.)



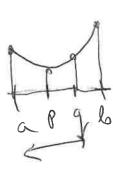
$$0.08 = M(0.2)^{\alpha}$$
 $(5.625 = 2.5^{\alpha})^{\alpha}$
 $0.005/2 = M(.08)^{\alpha}$ $(\alpha = 3.00)^{\alpha}$

3. If Newton's method is used to find a root of $f(x) = x^2 - R$, find bounds on x_0 for which convergence to the root \sqrt{R} is guaranteed. (Hint: for Newton's method, $e_{n+1} = \frac{1}{2} [f''(\psi_n)/f'(x_n)]e_n^2$, where ψ_n is between x_n and the root.)

$$C_{MH} = \frac{1}{2} \frac{2}{2x_n} C_n^2 = \left[\frac{x_n - J_R}{2x_n}\right] C_n$$

$$\left[\frac{x_n - J_R}{2x_n}\right] < 1 \quad \text{if} \quad \left[\frac{x_n - J_R}{2x_n}\right] < 1 \quad \text{or} \quad \left[\frac{x_n - J_R}{2x_$$

4. The golden search method tries to minimize f(x), where f is assumed to be unimodal in $a \le x \le b$, by evaluating f at two points between a and b, p = a + (1 - r) * (b - a) and q = a + r * (b - a), where r = 0.618... If f(q) is larger than f(p), the minimum is known to be in the new interval [a, q], otherwise the minimum is known to be in [p, b]. Why? Would this algorithm still work if we used r = 0.75? What is the advantage of using r = 0.618...?



If f(g) > f(p) the minimum cannot be in (g, b) otherwise f would be increasing for some points to the left of g and decreasing for some points to the right, hence not unimodal. Would still work with r=0.75 but would require two new function evaluation per iteration.

With $r^2=/-r$, can we old g ar new g in (a,g)

5. Write out the equations used to solve the following system using Newton's method:

$$f(x,y) = 1 + x^{2} - y^{2} + e^{x}\cos(y)$$

$$g(x,y) = 2xy + e^{x}\sin(y)$$

6. To solve $x^2 - 3x - 4 = 0$ we could write $x^2 = 3x + 4$, then x = 3 + 4/x, and iterate with this last formula: $x_{n+1} = 3+4/x_n$. Determine (without actually iterating) if this iteration will converge if we start near the root r=4. What if we start near the root r=-1?

$$X_{n+1} = 3 + \frac{4}{x_n} = g(x_n)$$

$$g'(x) = -\frac{4}{x^2} \quad g'(4) = -0.25 \quad \text{conveyor}$$

$$g'(-1) = -4 \quad \text{diverger}$$

Math 5329, Test I (k)

Name ____Key____

1. a. Write the Taylor polynomial $T_n(x)$ of degree n for the function f(x) = cos(x), expanded around a = 0.

$$T_{h}(x) = 1 - \frac{x^{2}}{2!} + \frac{x^{4}}{4!} - \frac{x^{6}}{6!} + \dots + \frac{x^{m}}{m!} (\pm 1)$$

$$m = even \pm 1$$

b. Find a reasonable upper bound on the error in $T_n(x)$ at x = 25 and estimate how big n needs to be for the error to be less than 10^{-3} .

c. Do you expect to have significant problems with roundoff error in calculating $T_n(25)$, with n as in part b? What if you calculate $T_n(1)$ with the same n?

$$y^{\text{ev}} \left(1 + \frac{x^{4}}{9!} + \frac{x^{8}}{8!} + \dots + \frac{x^{68}}{68!} \right) - \left(\frac{x^{2}}{2} + \frac{x^{4}}{6!} + \frac{x^{10}}{10!} + \frac{x^{70}}{70!} \right)$$

$$= \log_{10} - \log_{10} = 0.9912$$

$$NO_{10} T_{10}(1)$$

2. Show that the iteration $x_{n+1} = x_n - \frac{f(x_n)}{q(x_n)}$ converges quadratically (at least) to the root r of f(x) = 0, if $\lim_{x \to r} q(x) = f'(r) \neq 0$.

$$g(\omega) = x - \frac{f(\omega)}{g(\omega)}$$

$$g'(\omega) = 1 - \frac{f'(\omega)}{g(\omega)} + f(\omega) \frac{g'(\omega)}{g(\omega)^2}$$

3. For a certain root finder (Muller's method) it can be shown that $\lim_{n\to\infty}\frac{e_{n+1}}{e_ne_{n-1}e_{n-2}}=M(\neq 0,\neq \infty)$. To estimate the order α of this method, assume $e_{n+1}=Ce_n^{\alpha}$, and $e_{n+1}=Me_ne_{n-1}e_{n-2}$. Find an equation satisfied by α , you need not actually find α .

$$e_{n} = Ce_{n-1}^{x} = C\left(Ce_{n-2}^{x}\right)^{x} = C^{1+x}e_{n-1}^{x^{2}} \qquad x=1.839$$

$$e_{n+1} = (e_{n})^{\frac{1}{x}} \qquad e_{n-2} = (e_{n})^{\frac{1}{x}} \qquad (f_{n})^{\frac{1}{x}} = x$$

$$e_{n+1} = Me_{n}\left(\frac{e_{n}}{e}\right)^{\frac{1}{x}}\left(\frac{e_{n}}{e^{1+x}}\right)^{\frac{1}{x}} = xe_{n}^{1+x}$$

$$e_{n+1} = Me_{n}\left(\frac{e_{n}}{e}\right)^{\frac{1}{x}}\left(\frac{e_{n}}{e^{1+x}}\right)^{\frac{1}{x}} = xe_{n}^{1+x}$$

4. To minimize the function $f(x,y) = 100(x^2-y)^2 + (1-x)^2$, set f_x and f_y equal to zero, and do one iteration of Newton's method, starting from (1,0) to solve this system of two equations and two unknowns. From (1,0), what is the direction of steepest descent?

$$f = 100 \times^{4} - 200 \times^{2} y + 100 y^{2} + 1 - 2x + x^{2}$$

$$f_{x} = 900 \times^{3} - 900 \times y + 2x - 2 = 0$$

$$f_{y} = -200 \times^{2} + 200 y = 0$$

- 7_
- 5. Explain how Newton's method could be used to compute A/B on a computer which only can add, subtract and multiply, but not divide.

$$f(x) = \frac{1}{x} - B$$

$$x_{nq} = x_n - \frac{1}{x_n} = 2x_n - Bx_n^2 \quad \text{Herefore fo}$$

$$f(x) = \frac{1}{x_n} - \frac{1}{x_n} = 2x_n - Bx_n^2 \quad \text{Herefore fo}$$

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$$f(x) = \frac{1}{x_n} - \frac{1}{x_n} = \frac{1}{x_n} -$$

6. If $f(x) = (x-r)^m$, show that the "modified" Newton's method $x_{n+1} = x_n - m \frac{f(x_n)}{f'(x_n)}$ will converge in a single iteration to the root r, regardless of the starting value x_0 . What would you predict would happen if this modified Newton method were applied to a more general function with a root of multiplicity m at r, that is to $f(x) = (x-r)^m h(x)$, where $h(r) \neq 0$? You can analyze the iteration using the techniques of section 3.4, or you can guess; but if you guess, it must be correct!

$$X_{n+1} = x_n - m \frac{(x_n - r)^m}{m(x_n - r)^{n-1}} = x_n - (x_n - r) = r$$

$$will converge quechasis and $(x_n - r)^m d(x_n)$

$$x_{n+1} = x_n - m \frac{(x_n - r)^m d(x_n)}{(x_n - r)^m d(x_n)} + m(x_n - r)^{n-1} d(x_n)$$

$$x_{n+1} = x_n - m \frac{(x_n - r)^m d(x_n)}{(x_n - r)^m d(x_n)} + m(x_n - r)^{n-1} d(x_n)$$

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$$x_n = x_n - m \frac{(x_n - r)^m d(x_n)}{(x_n - r)^m d(x_n)} + m(x_n - r)^{n-1} d(x_n)$$

$$x_n = x_n - m \frac{(x_n - r)^m d(x_n)}{(x_n - r)^m d(x_n)} + m(x_n - r)^{n-1} d(x_n)$$

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$$x_n = x_n - m \frac{(x_n - r)^m d(x_n)}{(x_n - r)^m d(x_n)} + m(x_n - r)^m d(x_n)$$

$$x_n = x_n - m \frac{(x_n - r)^m d(x_n)}{(x_$$$$

 $g'(x) = 1 - m(x-r) \left(\frac{1}{2} - m(x-r) \frac{1}{2} + m(x) \right)$





a. Find $T_n(x)$, the Taylor series of degree n for the function f(x) = cosh(x), expanded around a = 0. Assume n is even.

(Hint: $\frac{d}{dx}cosh(x) = sinh(x)$, $\frac{d}{dx}sinh(x) = cosh(x)$, sinh(0)=0, cosh(0)=1)

$$T_{\Lambda}(x) = 1 + \frac{x^2}{2!} + \frac{x^4}{4!} + \frac{x^6}{6!} + \frac{x^{4}}{4!}$$

b. Find $E_n(x)$, the error in $T_n(x)$, and find a reasonable upper bound on $E_n(10)$. You can use the fact that sinh(x) is a monotone increasing function.

$$E_{n}(x) = \frac{f^{n+1}(\epsilon)}{(n+1)!} \times^{n+1} = \frac{57nh(\epsilon)}{(n+1)!} \times^{n+1}$$

$$|E_{n}(10)| \leq \frac{5nh(10)}{(n+1)!} / 0^{n+1}$$

c. Estimate the number of terms n required for $T_n(10)$ to approximate $\cosh(10)$ to an accuracy of 10^{-4} .

d. Would you expect roundoff error to be a serious concern in computing $T_n(10)$ in part (c)? Why or why not?

2.

tives. (XAH) = (Xn) - (Fxx Fxy) (Fx) (Fyx Fyy) (Fx)

a. To find a maximum or minimum of a function F(x,y), in calculus

we set both partial derivatives to 0 and solve the resulting system of two equations. Explicitly write out what Newton's method looks like when applied to this system, in terms of F and its deriva-

- 7
- b. If F(x,y) is a quadratic polynomial (F(x,y) = a + bx + cy + b) $dx^2 + exy + fy^2$), what can you say about convergence of Newton's method?

converger in a stryle interation

3. It can be shown that for Mueller's method, $e_{n+1} \approx Me_n e_{n-1} e_{n-2}$. If Mueller's method is order α , ie, $e_{n+1} \approx Ce_n^{\alpha}$, find an equation satisfied by α . Then use any method we have studied to find a root of this equation. (Hint: First write e_{n-1} and e_{n-2} in terms of e_n .)

$$e_{n} = ce_{n}^{x} e_{n} = (e_{n})^{x}$$

$$e_{n} = ce_{n}^{x} e_{n} = (e_{n})^{x} = De_{n}^{x}$$

$$e_{n+1} = Me_{n} e_{n}^{x} e_{n}^{x} = Me_{n}^{x} e_{n}^{x} = Me_{n}^{x} e_{n}^{x}$$

$$e_{n+1} = Me_{n}^{x} e_{n}^{x} e_{n}^{x} = Me_{n}^{x} e_{n}^{x} = Me_{n}^{x} e_{n}^{x}$$

4. About how many bisection iterations should be required to obtain an error less than ϵ , knowing that f(a) and f(b) have opposite signs?

 $\frac{b-a}{2^{n}} = \epsilon \qquad N = \frac{h\left(\frac{b-a}{\epsilon}\right)}{h 2}$



- 5. Estimate the order of convergence for:
 - a. Newton's method applied to $f(x) = (x-3)^3(x-4)$, starting near the root r=3.
 - b. Same as (a) but starting near the root r=4. 2
 - c. Same as (a) but using Secant method. /
 - d. Same as (a) but using Secant method and starting near the root r=4.
 - e. A root finder which produces consecutive errors of 10^{-5} , 10^{-7} and 10^{-12} . **2.5**
 - f. The iteration $x_{n+1} = g(x_n)$ if r = g(r) and g'(r) = g''(r) = 0 but $g'''(r) \neq 0$, and you start near the root r.
 - g. The bisection method.
 - h. The method $x_{n+1} = x_n f(x_n)/f'(x_0)$.

Name _______

1. a. Find $T_n(x)$, the Taylor series of degree n for the function f(x) = ln(1+x), expanded around a = 0.

(Hint: $f^{(n)}(x) = (-1)^{n-1}(n-1)!/(1+x)^n$, for $n \ge 1$.)

$$T_{\Lambda}(k) = \chi - \frac{\chi^2}{k} + \frac{\chi^3}{3} - \frac{\chi^4}{4} = \frac{\chi^4}{5} + \frac{\chi^4}{5} = \frac$$

b. Find $E_n(x)$, the error in $T_n(x)$, and find a reasonable upper bound on $E_n(1)$.

$$|E_{n}(x)| = \frac{|E_{n}(x)|^{2}}{(1+4)^{n+1}} \frac{|E_{n}(x)|^{2}}{(1+4)^{n+1}} = \frac{1}{(1+4)^{n+1}} \frac{|E_{n}(x)|^{2}}{(1+4)^{n+1}} = \frac{1}{(1+4)^{n+1}} \frac{|E_{n}(x)|^{2}}{(1+4)^{n+1}}$$

c. Estimate the number of terms n required for $T_n(x)$ to approximate $f(1) = \ln(2)$ to 5 decimal places accuracy.

d. Would you expect roundoff error to be a serious concern in (c)? Why or why not? (Hint: $1+1/2+1/3+1/4+1/5+...+1/n \approx ln(n)$, for large n.)

2. Estimate the order of convergence of a root-finder that has consecutive errors 0.2, 0.08, 0.002048.

39,0625 = 2.5×

3. A certain computer stores floating point numbers in a 128-bit word. If a floating point number is written in normalized binary form (1.xxxxx...₂* 2^e), it is stored using one sign bit (0 if the number is positive), then e + 4095 is stored in binary in the next 13 bits, and then the mantissa xxxxx... is stored in the final 114 bits. Show exactly how the number -27.125 would be stored on this computer. Also: approximately how many decimal digits of accuracy does this machine have?

Let:
$$1/1000000000011/101100100011$$

COOE C80...0 $2^{-1/4} = 5.10^{-35}$ so = 35 learner digits

4. The fixed point iteration $x_{n+1} = x_n + \sin(x_n)$ has roots at $r = n\pi$ for any integer n. Will this iteration converge if you start very close to the root r = 0? Will it converge if you start near the root $r = \pi$? In both cases, if it does converge, what is the order of convergence?

$$g(k) = x + 5n(k)$$

$$g'(k) = 1 + cw(k)$$

$$g''(x) = -5n(k)$$

$$g'''(x) = -cw(k)$$

diverger
$$g'(t) = 0$$
 $g'(t) = 0$
 $g''(t) = 0$
 $g'''(t) = 0$
 $g'''(t) = 0$
 $g'''(t) = 0$

5. Write the secant iteration for solving f(x) = 1/x - b = 0, in a form where no divisions are required (thus this iteration could be used to compute 1/b on a computer which cannot do divisions).

$$\begin{array}{lll}
X_{h+1} &= X_{h} & - & \frac{F(x_{h})(x_{h} - x_{h-1})}{F(x_{h}) - F(x_{h-1})} \\
&= X_{h} & - & \left(\frac{1}{X_{h}} - B\right)(x_{h} - X_{h-1}) & - & \left(\frac{1}{X_{h}} - BX_{h} - X_{h-1}\right) \\
&= \left(\frac{1}{X_{h}} - \frac{1}{X_{h}}\right) & - & \left(\frac{1}{X_{h}} - BX_{h} - X_{h-1}\right)
\end{array}$$

6. To minimize the function $f(x,y) = 10(2x+y)^2 + (x-2)^2$, set f_x and f_y equal to zero, and do one iteration of Newton's method, starting from (0,1) to solve this system of two equations and two unknowns. The true minimum is obvious from looking at the function, where is the minimum? From (0,1), what is the direction of steepest descent? Which converges faster, Newton's method or the method of steepest descent?

descent?

$$f_{x} = 20(2x+y)2 + 2(x-2) = 82x + 40y-4$$
 $f_{y} = 20(2x+y) = 40x + 20y$
 $Vf(0,1) = (36,20)$
 $\begin{pmatrix} x_{n+1} \\ y_{n+1} \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} - \begin{pmatrix} 82 & 40 \\ 40 & 20 \end{pmatrix} - \begin{pmatrix} 36 \\ 20 \end{pmatrix} = \begin{pmatrix} 2 \\ -4 \end{pmatrix}$
 $Vf(0,1) = (36,20)$
 $Vf(0,1) = (36,20)$

(on iteration !)