

1.8 bounding the error of expressions

modelling expressions with simple bivariate functions

- ▶ let a set of integers i_1, \dots, i_n and j_1, \dots, j_n satisfy
 - ▶ either $i_k = j_k = 0$
 - ▶ or $j_k < i_k < k$
- ▶ let f_1, \dots, f_n be bivariate real functions defined on compact domains
 - ▶ the functions f_k are either arithmetic binary operations or univariate functions
- ▶ let $u_0 = 0$ and u_k be defined by the system of equations

$$u_k = f_k(u_{i_k}, u_{j_k}), \quad k = 1, \dots, n$$

evaluation of the expression

- ▶ these equations are thus solved (i.e. all u_k computed) by substitution

$$u_1 = f_1(u_0, u_0) = f_1(0, 0)$$

$$u_2 = f_2(u_{i_2}, u_0) = f_2(u_{i_2}, 0), \quad i_2 \in \{0, 1\}$$

$$u_3 = f_3(u_{i_3}, u_{j_3}), \quad i_3 \in \{0, 1, 2\}, j_3 \in \{0, \dots, i_3\}$$

...

$$u_n = f_n(u_{i_n}, u_{j_n}), \quad i_n \in \{0, \dots, n-1\}, j_n \in \{0, \dots, i_n\}$$

- ▶ with this we have modeled the evaluation of numerical expressions where u_n is the value of the expression and the other u_k intermediate results

example $\left(-p + \sqrt{p^2 - 4q}\right) / 2$

$$u_1 = p$$

$$u_2 = q$$

$$u_3 = u_1^2$$

$$u_4 = u_3 - 4u_2$$

$$u_5 = \sqrt{u_4}$$

$$u_6 = (-u_1 + u_5) / 2$$

the same with rounding errors at every step

- ▶ now let v_k be the numerical versions of u_k defined by

$$v_k = (1 + \delta_k) f_k(v_{i_k}, v_{j_k}), \quad k = 1, \dots, n$$

and $v_0 = 0$

- ▶ as usual $|\delta_k| \leq \epsilon$
- ▶ the relative error of v_k , i.e., $(v_k - u_k)/u_k$ is denoted by θ_k so that

$$v_k = (1 + \theta_k)u_k$$

example with rounding errors

$$v_1 = (1 + \delta_1)p$$

$$v_2 = (1 + \delta_2)q$$

$$v_3 = (1 + \delta_3)v_1^2$$

$$v_4 = (1 + \delta_4)(v_3 - 4 v_2)$$

$$v_5 = (1 + \delta_5)\sqrt{v_4}$$

$$v_6 = (1 + \delta_6)(-v_1 + v_5)/2$$

total error at every step – for multiplication and division

- ▶ recall: $f_k(x_i, x_j)$ is either an arithmetic binary operation (like sum) of x_i and x_j or a unary operation $f_k(x_i)$
- ▶ the simplest cases are multiplication and division
- ▶ for multiplication $f_k(v_i, v_j) = (1 + \theta_i)(1 + \theta_j)u_i u_j$ and so

$$v_k = (1 + \delta_k)(1 + \theta_i)(1 + \theta_j) u_k$$

- ▶ multiplication:

$$\theta_k = (1 + \delta_k)(1 + \theta_i)(1 + \theta_j) - 1 \approx \theta_i + \theta_j + \delta_k$$

- ▶ division:

$$\theta_k = (1 + \delta_k)(1 + \theta_i)/(1 + \theta_j) - 1 \approx \theta_i - \theta_j + \delta_k$$

total error at every step – for addition and subtraction

- ▶ for addition $f_k(v_i, v_j) = (1 + \theta_i)u_i + (1 + \theta_j)u_j$ and so

$$v_k = (1 + \delta_k) \left((1 + \theta_i) \frac{u_i}{u_i + u_j} + (1 + \theta_j) \frac{u_j}{u_i + u_j} \right) (u_i + u_j)$$

- ▶ addition:

$$\theta_k = (1 + \delta_k) (1 + \zeta_k \theta_i + (1 - \zeta_k) \theta_j) - 1 \approx \zeta_k \theta_i + (1 - \zeta_k) \theta_j + \delta_k$$

where $\zeta_k = u_i / (u_i + u_j)$

- ▶ convex combination if u_i and u_j have equal sign
- ▶ if different sign, error can be very large despite the fact that some times $\delta_k = 0$ in this case
- ▶ similar for subtraction

total error at every step – for univariate function

- ▶ $f_k(v_i) = f_k((1 + \theta_i)u_i)$ and so

$$\begin{aligned}v_k &= (1 + \delta_k) f_k((1 + \theta_i)u_i) \\&= (1 + \delta_k) \left(1 + \frac{f_k((1 + \theta_i)u_i) - f_k(u_i)}{f_k(u_i)} \right) u_k \\&= (1 + \delta_k) (1 + \zeta_k \theta_i) u_k\end{aligned}$$

where $\zeta_k = \frac{f_k((1+\theta_i)u_i) - f_k(u_i)}{f_k(u_i)}$ and

$$|\zeta_k| \leq \frac{L_k |u_i|}{|f(u_i)|}$$

if L_k is Lipschitz constant of f_k

- ▶ relative error of v_k is then

$$\theta_k = (1 + \delta_k)(1 + \zeta_k \theta_i) - 1 \approx \zeta_k \theta_i + \delta_k$$

relative errors for example

$$\theta_1 = \delta_1$$

$$\theta_2 = \delta_2$$

$$\theta_3 = (1 + \delta_3)(1 + \theta_1)^2 - 1$$

$$\theta_4 = (1 + \delta_4)(1 + \zeta_4\theta_3 - (1 - \zeta_4)\theta_2) - 1$$

$$\theta_5 = (1 + \delta_5)(1 + \zeta_5\theta_4) - 1$$

$$\theta_6 = (1 + \delta_6)(1 - \zeta_6\theta_1 + (1 - \zeta_6)\theta_5) - 1$$

- homework: what are the ζ_k , get bounds and obtain a bound for θ_6

stability and growth factor

- ▶ we say that the f_k are **stable** for if there exists some $L > 0$ such that for all k one has

$$|f_k(x_1, x_2) - f_k(y_1, y_2)| \leq L \max_i |x_i - y_i|$$

- ▶ we assume that for $k > 0$ one has $u_k \neq 0$
- ▶ then one can define a *growth factor*

$$\rho = \max\{|u_j|/|u_k| \mid j < k\}$$

a simple global error bound

Proposition Let $\alpha = (1 + \epsilon)L\rho$ where L be as defined above, ρ be the growth factor then

$$v_k = (1 + \theta_k)u_k$$

where

$$|\theta_k| \leq \left(\frac{\alpha^{k+1} - 1}{\alpha - 1} \right) \epsilon$$

proof.

- ▶ induction
- ▶ first one has

$$v_1 = (1 + \delta_1)u_1$$

and thus $\theta_1 = \delta_1$ and $|\theta_1| = |\delta_1| \leq \epsilon$

- ▶ then

$$\begin{aligned} v_{k+1} &= (1 + \delta_{k+1})f_{k+1}(v_{i_{k+1}}, v_{j_{k+1}}) \\ &= (1 + \theta_{k+1})u_{k+1} \end{aligned}$$

where

$$\theta_{k+1} = \delta_{k+1} + (1 + \delta_{k+1}) \frac{f_{k+1}(v_{i_{k+1}}, v_{j_{k+1}}) - f_{k+1}(u_{i_{k+1}}, u_{j_{k+1}})}{u_{k+1}}$$

- ▶ the (absolute value of the) first term is bounded by ϵ and for the second term one has for some $0 < i \leq k$:

$$\begin{aligned}
 (1 + \delta_{k+1}) \left| \frac{f_{k+1}(v_{i_{k+1}}, v_{j_{k+1}}) - f_{k+1}(u_{i_{k+1}}, u_{j_{k+1}})}{u_{k+1}} \right| &\leq (1 + \epsilon) L \frac{|v_i - u_i|}{|u_{k+1}|} \\
 &= \frac{(1 + \epsilon) L |\theta_i| \cdot |u_i|}{|u_{k+1}|} \\
 &\leq L(1 + \epsilon) \frac{\alpha^{i+1} - 1}{\alpha - 1} \\
 &\leq \frac{\alpha^{k+2} - \alpha}{\alpha - 1} \epsilon
 \end{aligned}$$

from which one gets

$$|\theta_{k+1}| \leq \frac{\alpha^{k+2} - 1}{\alpha - 1} \epsilon$$



example: graph

Math3511

Tuesday, 6 March 2018 11:07 AM

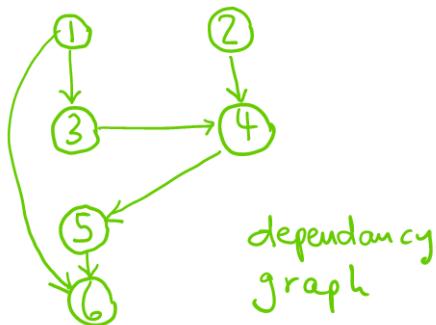


Figure 1: Graph.png