

## INTRODUCTION

- Floating point arithmetic susceptible to rounding errors - loss of accuracy.
- Dynamic error analysis an effective tool for measuring sensitivity.
- Implementation difficult - requires significant modifications to existing source code.

## AIMS

- Automate the quantitative analysis of floating point rounding errors.
- Improved methods for the analysis of results.
- Demonstrate applications.

## BACKGROUND

### Floating Point Arithmetic:

- Binary IEEE754 Floating Point - subset of real numbers:

$$x = (-1)^s m \beta^e$$

- Implemented as finite precision rounded arithmetic system. Exact values are rounded approximations of inexact values:

$$\begin{aligned} \hat{x} &= \mathbb{F}(x) \\ &= x(1 + \delta) \end{aligned}$$

- Normalization stage can lead to cancellation

### Monte Carlo Arithmetic:

- Track information lost - model inexactness using random perturbations:

$$\begin{aligned} \text{inexact}(x, t, \xi) &= x + 2^{e_x - t} \xi \\ &= (-1)^{s_x} (m_x + 2^{-t} \xi) 2^{e_x} \end{aligned}$$

- Statistics on rounding error obtained through repeated computations
- Uniformly distributed random variable used for  $\xi$
- Individual operators performed in terms of the inexact function:

$$x \circ y = \text{round}(\text{inexact}(\text{inexact}(x) \circ \text{inexact}(y)))$$

## LIBRARY IMPLEMENTATION

- Operations converted with source to source compiler CILLY
- Generated file compiled with MCA library to produce executable:

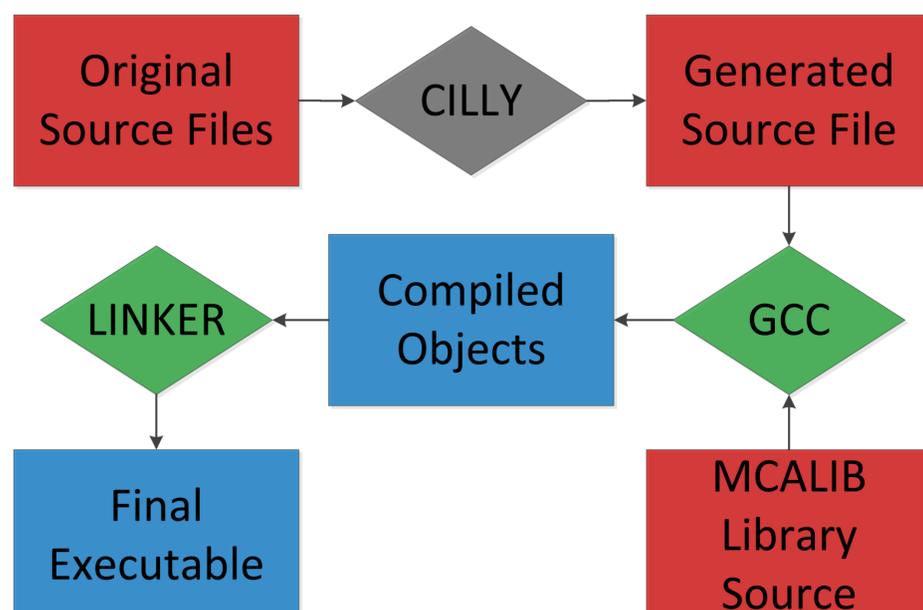


Figure 1: Source to Source compilation and Library implementation

## METHODS FOR ANALYSIS OF RESULTS

- Results analysis in previous publications limited
- Approach may be more formally defined - provide better interpretation of MCA results
- Approach defines sensitivity to rounding error with two measurements:
  - The number of base-2 digits lost to rounding error,  $K$
  - The minimum precision required to avoid a complete loss of significance,  $t_{min}$
- Measurements are found by performing linear regression using relative standard deviation,  $(\Theta)$ , as the dependant variable and,  $t$ , as the exploratory variable:

$$\begin{aligned} \log_{10}(\Theta) &= \log_{10}(2^{K-t}) \\ &= -\log_{10}(2)t + \log_{10}(2)K \\ &= mt + c \end{aligned}$$

- Robust regression methods are used to detect outlying data points
- Slope is a known variable - reduces problem to a 1D optimization

## TESTING & RESULTS

- Statistical results and new analysis techniques allow for the comparison of algorithms and detection of catastrophic cancellation:

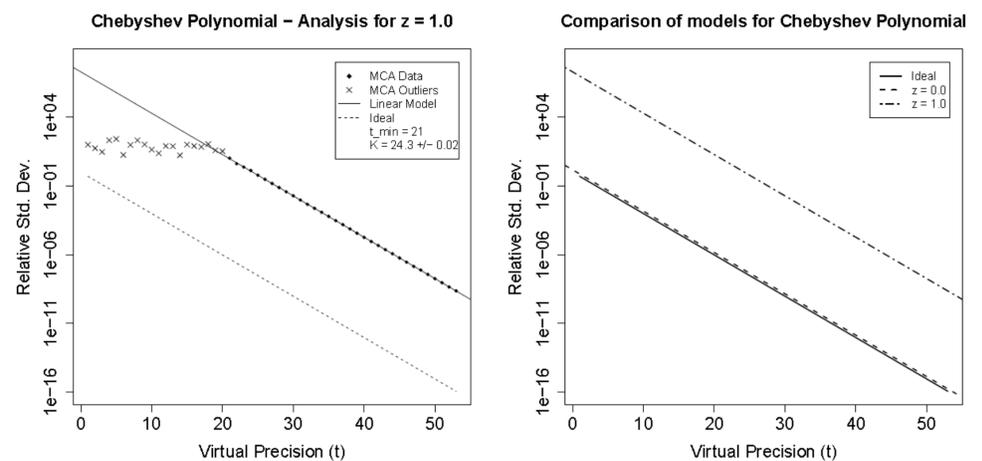


Figure 2: Comparison of results for Chebyshev Polynomial

- Analysis of results allows for optimization of algorithms:

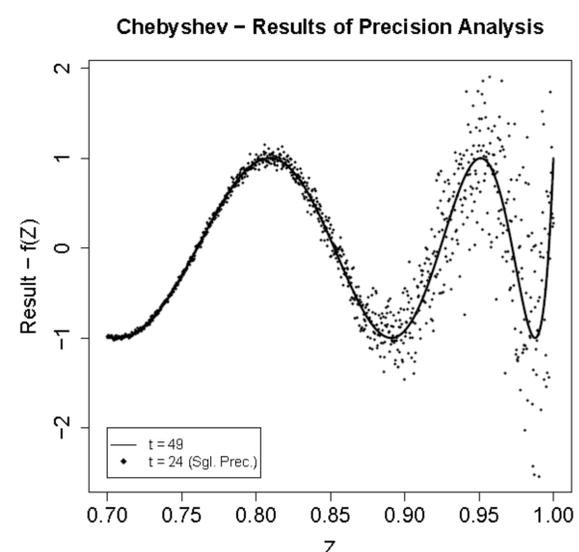


Figure 3: Optimization of Chebyshev Polynomial

## CONCLUSIONS

- No automated methods for measuring sensitivity to rounding error available
- MCALIB – determine if single or double precision floating point arithmetic is required, compare implementations or optimize software for precision.
- First implementation of it's type to perform these functions using automated dynamic analysis methods.
- Represents a revolution in the field of error analysis.