

Analyzing Fixed Point Arithmetic Rounding Error

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Abstract

We examine the magnitude of rounding error resulting from different combinations of fixed point arithmetic operations used to achieve the same goal, in an environment with truncating integer semantics.

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1 Definitions

1.0.1 Epsilon Notation

By definition of the floor operation, the following is true:

$$\frac{x}{y} - 1 < \left\lfloor \frac{x}{y} \right\rfloor \leq \frac{x}{y}$$

To simplify the inequality above for use in larger expressions, we will define epsilon (ϵ) as a non-deterministic number in the interval $[0, 1)$. Using this new notation, we can rewrite the inequality above like so:

$$\frac{x}{y} - \epsilon = \left\lfloor \frac{x}{y} \right\rfloor$$

2 Analysis

A common use case for fixed point arithmetic is multiplying a quantity of one asset by a conversion rate to another asset, in a programming language with no floating point arithmetic, only integer operations. A notable example of a language with these properties is the Solidity smart contract language, which only has basic truncating integer operations.

2.1 Introduction

Let σ_n be an arbitrary quantity of asset A that we wish to exchange for a quantity of asset B, μ_n . The rate of exchange between asset A and asset B is defined as the ratio between the separate quantities of μ and σ .

Below we will define and analyze multiple implementations of the desired function, $\text{to}_\mu(\sigma_n)$, which takes a quantity of asset A (σ_n) and returns the amount of asset B that the quantity of asset A is worth (μ_n) using the rate of exchange formula described briefly above ($\frac{\mu}{\sigma}$).

2.2 Fixed Loss Implementation

In an environment without truncating division, we can observe that these two implementations are equivalent:

$$\mu_n = \text{to}_\mu(\sigma_n) = \sigma_n \cdot \frac{\mu}{\sigma} = \frac{\sigma_n \mu}{\sigma}$$

However, in an environment with truncating division, these implementations differ:

$$\mu_n = \text{to}_\mu(\sigma_n) = \sigma_n \cdot \left\lfloor \frac{\mu}{\sigma} \right\rfloor \leq \left\lfloor \frac{\sigma_n \mu}{\sigma} \right\rfloor$$

Rewriting using the epsilon notation introduced earlier:

$$\mu_n = \text{to}_\mu(\sigma_n) = \sigma_n \cdot \left(\frac{\mu}{\sigma} - \epsilon \right) \leq \frac{\sigma_n \mu}{\sigma} - \epsilon$$

Distributing multiplication:

$$\mu_n = \text{to}_\mu(\sigma_n) = \frac{\mu}{\sigma} \sigma_n - \epsilon \sigma_n \leq \frac{\sigma_n \mu}{\sigma} - \epsilon$$

Now, with the expressions fully expanded, we can easily observe that the expression where multiplication is performed before division is a much better approximation of the pure, non-truncating exchange rate formula.

Compared to the scaled rounding loss in the left equation ($\epsilon \sigma_n$), where rounding error is unbounded, rounding error in the equation on the right is bounded between $[0, 1)$.

2.3 Fixed Point Arithmetic Implementations

2.3.1 $\sigma_n.fmul(\mu.fdiv(\sigma))$

$$\mu_n = \text{to}_\mu(\sigma_n) = \left\lfloor \frac{\left\lfloor \frac{\mu \cdot 10^{18}}{\sigma} \right\rfloor \cdot \sigma_n}{10^{18}} \right\rfloor$$

Using epsilon notation:

$$\mu_n = \text{to}_\mu(\sigma_n) = \frac{(\frac{\mu \cdot 10^{18}}{\sigma} - \epsilon) \cdot \sigma_n}{10^{18}} - \epsilon$$

Simplified:

$$\mu_n = \text{to}_\mu(\sigma_n) = \frac{\mu \sigma_n}{\sigma} - \frac{\epsilon \sigma_n}{10^{18}} - \epsilon$$

Error scales proportionally with σ_n .

2.3.2 $\sigma_n.fmul(\mu).fdiv(\sigma)$

$$\mu_n = \text{to}_\mu(\sigma_n) = \left\lfloor \frac{\left\lfloor \frac{\sigma_n \cdot \mu}{10^{18}} \right\rfloor \cdot 10^{18}}{\sigma} \right\rfloor$$

Epsilon notation:

$$\mu_n = \text{to}_\mu(\sigma_n) = \frac{(\frac{\sigma_n \cdot \mu}{10^{18}} - \epsilon) \cdot 10^{18}}{\sigma} - \epsilon$$

Simplified:

$$\mu_n = \text{to}_\mu(\sigma_n) = \frac{\mu \sigma_n}{\sigma} - \frac{\epsilon 10^{18}}{\sigma} - \epsilon$$

Error is inversely proportional with σ .

2.3.3 $\sigma_n.fdiv(\sigma).fmul(\mu)$

$$\mu_n = \text{to}_\mu(\sigma_n) = \left\lfloor \frac{\left\lfloor \frac{\sigma_n \cdot 10^{18}}{\sigma} \right\rfloor \cdot \mu}{10^{18}} \right\rfloor$$

Epsilon notation:

$$\mu_n = \text{to}_\mu(\sigma_n) = \frac{(\frac{\sigma_n \cdot 10^{18}}{\sigma} - \epsilon) \cdot \mu}{10^{18}} - \epsilon$$

Simplified:

$$\mu_n = \text{to}_\mu(\sigma_n) = \frac{\mu \sigma_n}{\sigma} - \frac{\epsilon \mu}{10^{18}} - \epsilon$$

Error scales proportionally with μ .