

## Beyond the Circle: Deforming Contours in Inverse Z-Transform

Final Year Project Report

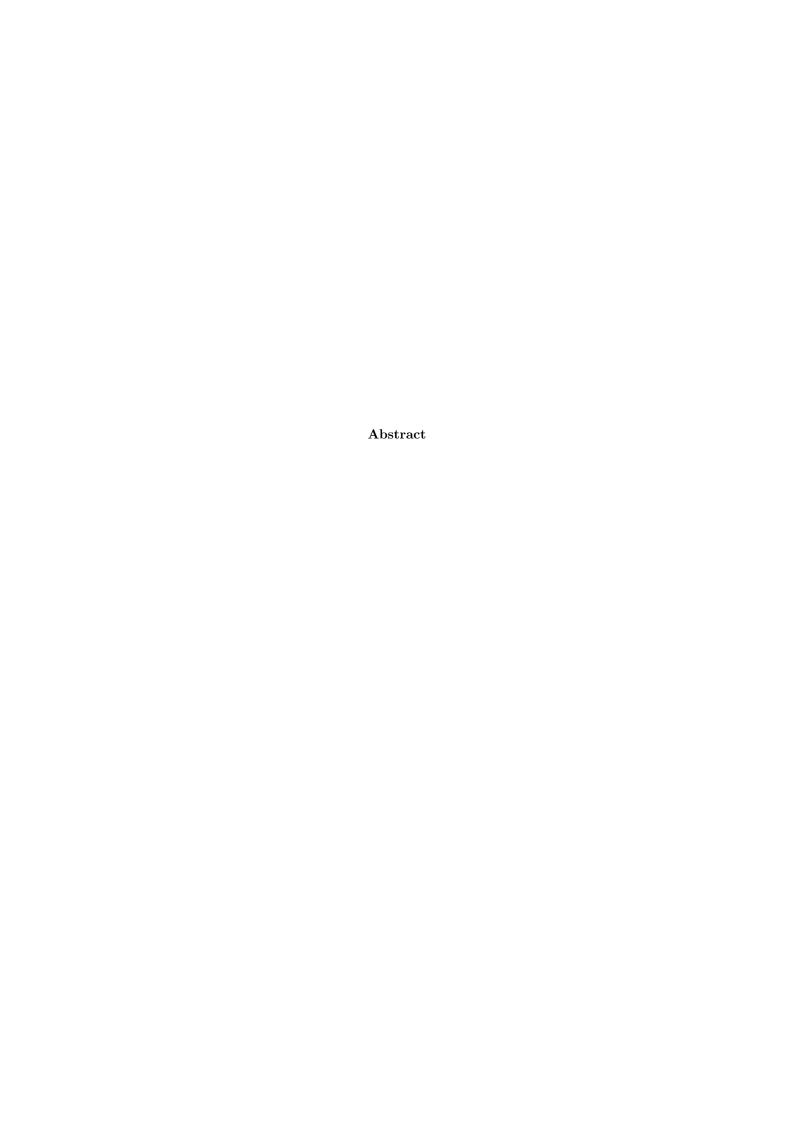
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<sup>1</sup>Disclaimer: This report is submitted as part requirement for the MEng degree in Mathematical Computation at UCL. It is substantially the result of my own work except where explicitly indicated in the text. The report may be freely copied and distributed provided the source is explicitly acknowledged.



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# Introduction

- 1.1 Motivation
- 1.2 Aims and Objectives
- 1.3 Overview

## Background

"By definition, a complex number z is an ordered pair (x,y) of real numbers x and y, written z = (x,y)" [Kreyszig, 2010]. In practice, complex numbers are written in the form z = x + iy, where x and y are real numbers and i is the imaginary unit. The set of complex numbers is denoted by  $\mathbb{C}$ .

#### 2.1 The $\mathcal{Z}$ -Transform

#### Come up with a more catchy Z-Transform title

The z-transform is a transformation of a real or complex continuous time function x(t), often used for discrete time signals and is commonly described as the discrete time Fourier transform (DTFT). Taking the Fourier transform of a sampled function results in

$$\mathcal{F}\left[x(t)\sum_{n=-\infty}^{\infty}\delta(t-n\triangle t)\right] = \int_{-\infty}^{\infty}x(t)\sum_{n=-\infty}^{\infty}\delta(t-n\triangle t)e^{-i\omega t}dt$$
 (2.1)

We can then make use of the sifting property of the delta function.

$$= \sum_{n=-\infty}^{\infty} \int_{-\infty}^{\infty} x(t)\delta(t - n\Delta t)e^{-i\omega t}dt$$
 (2.2)

$$=\sum_{n=-\infty}^{\infty} x(n\triangle t)e^{-i\omega nt}$$
(2.3)

If we normalise the sampling interval to 1, we get

$$\sum_{n=-\infty}^{\infty} x(n)e^{-in\omega} \tag{2.4}$$

The sequence x(n) is sampled at discrete time intervals  $t_n = n \triangle t$ , where the sampling interval  $\triangle t$  is the time between consecutive samples and the time index n numbers the samples. The DTFT is a periodic function of  $\omega$  with period  $2\pi$ . The existence of the DTFT relies on the condition of absolute summability of the sequence x(n); all the terms must converge to a finite value. This requirement is mathematically expressed as

$$\sum_{n=-\infty}^{\infty} |x(n)| < \infty \tag{2.5}$$

We can extend our analysis to the Z-transform, which generalizes the discrete time Fourier transform to the complex plane, not just the unit circle where r = 1. The Z-transform of a sequence x(n) is formally given by

$$X(z) = \mathcal{Z}_{n \to z}[x(n)] = \sum_{n = -\infty}^{\infty} x(n)z^{-n}$$

$$(2.6)$$

For a complete description of z in the complex plane, we tend to its polar form  $z = re^{i\theta}$ , where r represents the magnitude of z and  $\theta$  (often written as  $\omega$  in the context of the unit circle for the DTFT) represents the angle of z with respect to the positive real axis.

In the analysis of causal systems - systems for which a time origin is defined and is illogical to consider signal values for negative time - the unilateral Z-transform is used. Unlike the bilateral Z-transform in (2.6), summing from zero to positive infinity yields the unilateral Z-transform defined to be

$$X(z) = \mathcal{Z}_{n \to z}[x(n)] = \sum_{n=0}^{\infty} x(n)z^{-n}$$
(2.7)

The region within the complex z-plane where (2.7) converges is known as the Region of Convergence (ROC). If we say that  $z_1$  converges, then  $z_1$  exists within the ROC. Thus, all z such that  $|z| > |z_1|$  also converge. Strictly speaking, the ROC is defined for the set of values of z for which the Z-transform is absolutely summable

$$\mathbf{ROC} = \left\{ z : \sum_{n=0}^{\infty} |x(n)z^{-n}| < \infty \right\}$$
 (2.8)

The ROC for causal sequences is typically the exterior of the outermost pole in the Z-plane, denoted as |z| > a. This region excludes the poles themselves, as the transform does not converge at those points. For the system to be *stable*, it's essential that the ROC includes the unit circle, |z| = 1. Thus, all poles must lie within the unit circle [Loveless and Germano, 2021].

#### 2.1.1 Probability Generating Functions

May be better to move within Abate and Whitt's method

#### 2.2 The Inverse $\mathcal{Z}$ -Transform

The inverse Z-transform aims to find the n-th value of the sequence x(n) given the Z-transform X(z). This is commonly defined as a Cauchy integral around a contour C in the complex plane. The contour C is a counter-clockwise closed path that encloses the region of convergence (ROC). The inverse Z-transform is formally given by

$$x(n) = \mathcal{Z}_{z \to n}^{-1}[X(z)] = \frac{1}{2\pi i} \oint_C X(z) z^{n-1} dz$$
 (2.9)

- 2.2.1 Abate and Whitt 1992
- 2.2.2 Cavers 1978
- 2.2.3 Series acceleration techniques
- 2.3 Discrete Pricing Options
- 2.3.1 Lookback and Barrier Options
- 2.4 Optimisation Techniques
- 2.4.1 Stochastic Gradient Descent
- 2.4.2 Bayesian Optimisation

# Experiment

Finding different parameters to use for the experiment making use of Machine Learning techniques.

# Results

# Conclusion

- 5.1 Summary
- 5.2 Future Work
- 5.3 Acknowledgements

# Bibliography

[Kreyszig, 2010] Kreyszig, E. (2010). Advanced Engineering Mathematics. John Wiley & Sons.

[Loveless and Germano, 2021] Loveless, B. and Germano, G. (2021). Review of numerical inversion techniques of the z-transform. Working Paper.

# Appendices

# Appendix A

# Initial Project Plan



# Numerical Benchmarking on Inverse Z-Transform and Its Uses in Discrete Pricing Options

Project Plan

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## Aims and Objectives

#### 1.1 Aims

We aim to understand a new efficient method for numerical evaluation of the inverse Z-transform, which states to be faster and more accurate than the standard trapezoid rule. A specific area of applying this method would be to the pricing of discretely monitored exotic options, such as lookback and barrier options, and see how it compares to other methods; Abate and Whitt's approach, C. Cavers' method with Euler, Shanks and epsilon accelerations, etc.

#### 1.2 Objectives

- Understanding Levendorskii's inverse Z-transform and the common numerical evaluation methods
- Implementing the function as a code
- Numerical benchmarking; average error, maximum error and CPU time
- Exploring its uses in discrete pricing options

#### 1.3 Deliverables

- numerical benchmarking results to add to 'Review of numerical inversion techniques of the z-transform' by Loveless and Germano
- results and implementation in regards to discrete pricing options (Accurate numerical inverse z-transform and it's use in the Fourier-z pricing of discretely monitored path-dependent options by Loveless, Phelan and Germano)

### Work Plan

### 2.1 Project Start $ightarrow 30^{ m th}$ November '23

- background reading on complex numbers & contour integration based methods, fourier transform, z-transform and its inverse, numerical approaches to inverse z-transform and pricing options (barrier and lookback options)
- $\bullet$  coding implementation of Levendorskii's inverse z-transform

### $2.2 \quad 1^{\mathrm{st}} \; \mathrm{December} \; `23 ightarrow 24^{\mathrm{th}} \; \mathrm{January} \; `24$

- preliminary research on Loveless' and Germano's 'Review of numerical inversion techniques of the z-transform'
- understanding the other methods; AW, C, CEuler, CShanks and CEpsilon
- going over the different functions; Heaviside Step, Polynomial, Decaying Exp, Sinusodial
- reviewing the code for numerical benchmarking
- implementing it for Levendorskii's method
- begin work on interim report

## 2.3 $24^{\rm th}$ January '24 $\rightarrow$ 15<sup>th</sup> March '24

- ullet preliminary recap on discrete pricing options (barrier and lookback options) and the need for z-transform
- use-case in discrete pricing options
- start work on project report; however, to be worked on throughout the year/stages

### $2.4~~5^{th}~March~`24 ightarrow 26^{th}~April~`24$

- extra time to deal with any unexpected problems or delays
- final touches