

Introduction

n-Widths, Total Positivity, and Chebyshev Systems

A Lower Bound for Worst-case Approximation Error

A Quartet of New Approximations

Examples

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Portfolio Approximation

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January 23, 2015

Notices

- The methods described here are patent pending.
- This work is my own; it does not necessarily reflect any views held or any approaches used by Convexity Capital Management LP.

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The Problem

- Find simple, accurate **approximations** of portfolio values across a range of scenarios (no model assumed)
- Consider only exponential-affine instruments
 - Bonds
 - Interest rate swaps
 - Credit default swaps
 - Equities (as streams of dividends)
- State of the art: duration and higher-order relatives
- **But** these only work well for scenarios with **small** changes
- I seek simple approximations accurate for both small and **large** changes

Do We Care?

- An (optimal) n -factor approximation yields an (optimal) n -instrument **approximating portfolio**
- An approximation that works only for **small** changes → accurate portfolio replication only for **small** moves
- An approximation that works even for **large** changes → accurate portfolio replication even in **large** moves

History of the Problem

- **Tables** of discount / growth factors: Trenchant (1558)
- **Continuous compounding**: Bernoulli (1690)
- **Duration** (Macaulay (1938), rediscovered and reinterpreted by Hicks (1939) and Samuelson (1945))
- **Immunization** (Redington (1952), higher-order Taylor series terms appear here, seemingly for the first time)
- **Model-based** approximation (Vasicek (1977), Cox *et al.* (1979), Cox *et al.* (1985), Heath *et al.* (1992))

A Quartet of New Approximations

- No model is assumed; I seek to minimize worst-case approximation error or average squared approximation error across a **range** of scenarios.
- Both small and large changes are handled.

	Worst-case Error	Avg. Squared Error
Portfolio	Numerically-optimal	Interpolatory
Nested	Mathieu	Oblate

Relations to Recent Literature

- Conceptually, worst-case approximation error is a focus when **robustness** of replication is a concern
 - ★ Hansen & Sargent (1995), Hansen *et al.* (1999), Anderson *et al.* (2003), Hansen *et al.* (2006), Hansen & Sargent (2008), through Hansen & Sargent (2012)
 - ★ Chamberlain (2000) and Chamberlain (2001)
- Technically, the average squared approximation error analysis involves classical extensions of **Perron-Frobenius** theory
 - ★ Hansen & Scheinkman (2009), Hansen (2012), and Borovička *et al.* (2014)
 - ★ Ross (2014)

Further Potential Applications

- I do **not** assume any model
- However, the approximations I develop can be applied directly to standard **affine models** and some variants
(Duffie *et al.* (2000), Piazzesi (2010), Campbell *et al.* (2013), Greenwood & Vayanos (2014))
- My approximations may also be useful in other contexts
(Campbell *et al.* (2014), Barndorff-Nielsen & Shephard (2001))

The Key Initial Step

- All problems of this type can be reduced to approximating one integral operator.
- The **exponential product operator** P_c takes a signed measure $h(x)$ (a cashflow stream) to a function of y (a scenario profile):

$$P_c [h] (y) = \int_{-1}^1 \exp(cxy) dh(x). \quad (1)$$

- $x \in [-1, 1]$ is the (normalized) scenario shift times time, while $y \in [-1, 1]$ is a fraction of the scenario change.
- c represents the product of half the range of scenario changes and half the range of “move times time.”

What Are the Domain and Range?

- P_c maps the Hilbert space $L^2[-1, 1]$ to itself.
- This domain and range work for average squared approximation error and smooth cashflow streams.
- P_c also maps signed measures of bounded variation (with the total-variation norm) to continuous functions (with the sup norm).
- This domain and range are necessary to handle worst-case approximation error and general cashflow streams.

Rank-*n* Approximations

- Seek approximations of the type

$$\exp(cx\mathbf{y}) \sim \sum_{i=1}^n f_i(x) g_i(y) \quad (2)$$

- What are the optimal functions f_i and g_i (optimal dimension reduction)?
- This is a problem in **linear *n*-widths**
- I provide methods for average squared error and **worst-case error**

Some Suboptimal Approximations

- Taylor series: $\exp(cx) \sim \sum_{i=1}^n \frac{c^{i-1}}{(i-1)!} x^{i-1} y^{i-1}$.
Worst-case error on the order of $1/n!$.
- Univariate Remez: $\exp(cx) \sim \sum_{i=1}^n a_i x^{i-1} y^{i-1}$.
Comes from applying the Remez method to
 $\exp(cu)$ as a function of $u \in [-1, 1]$.
Worst-case error on the order of $1/(2^{n-1} n!)$.
- How far are these from optimal? Very far indeed!

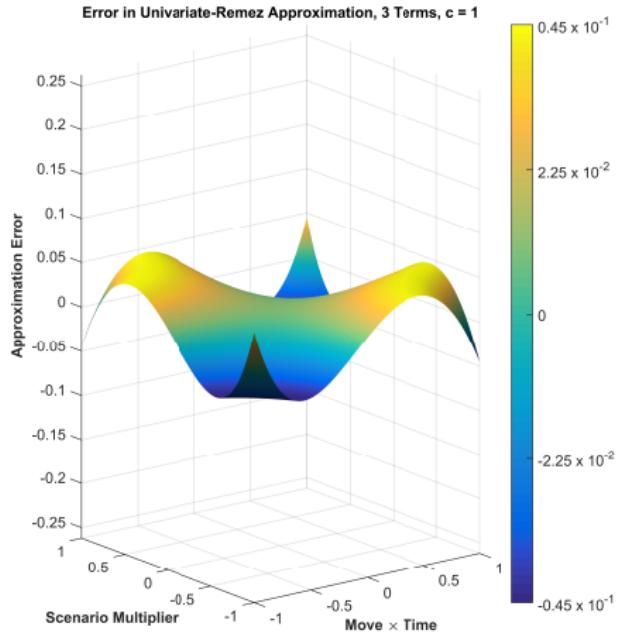
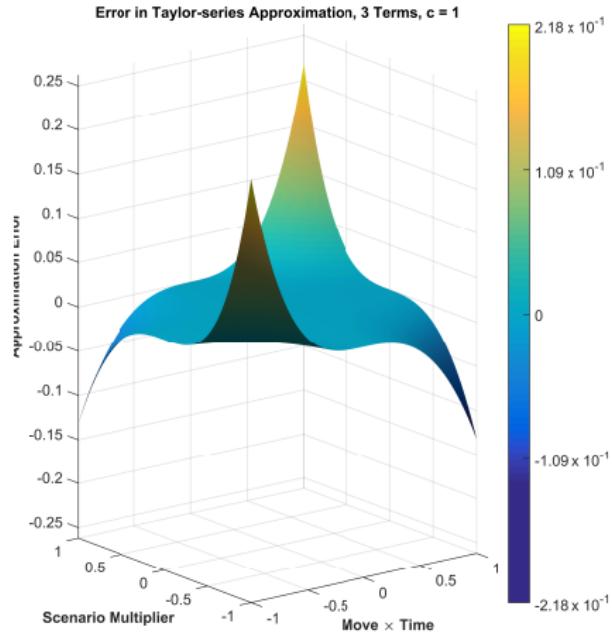
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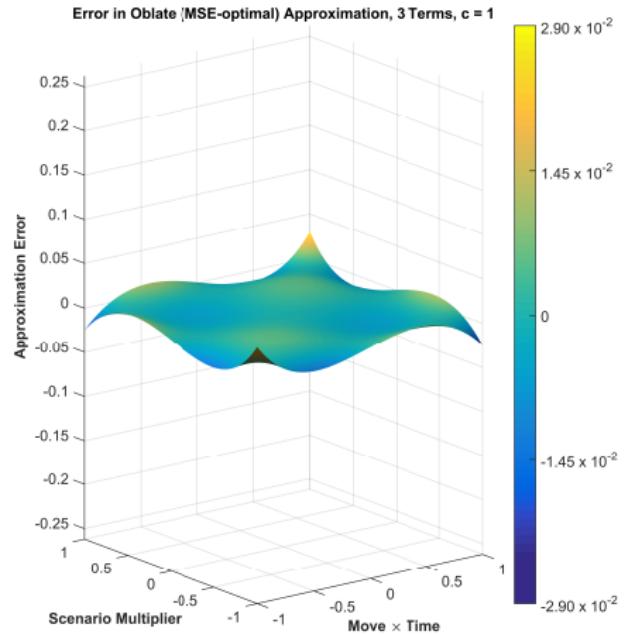
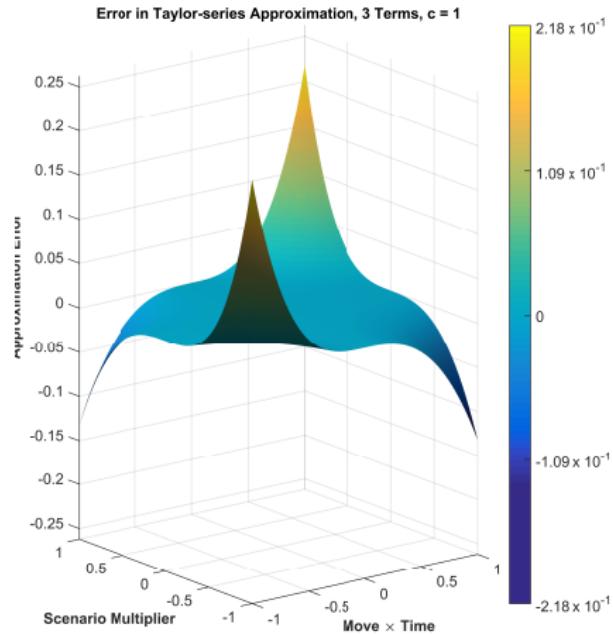
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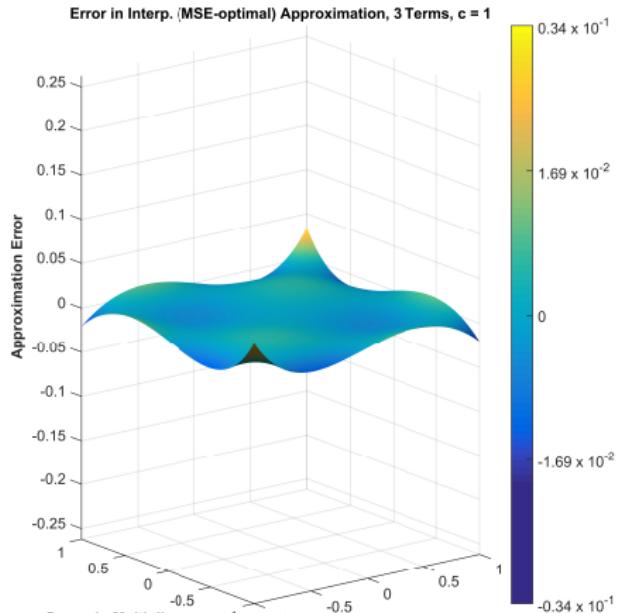
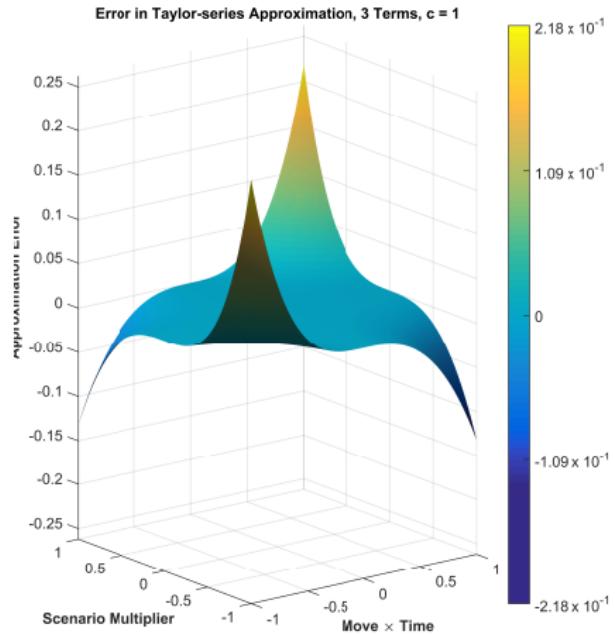
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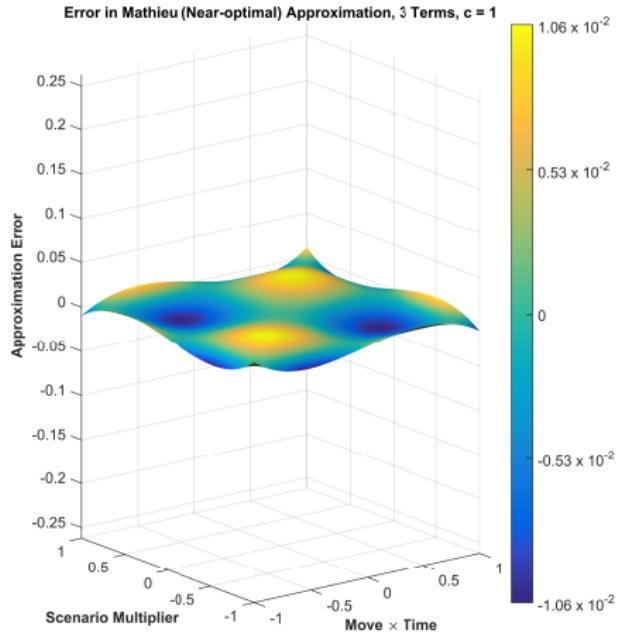
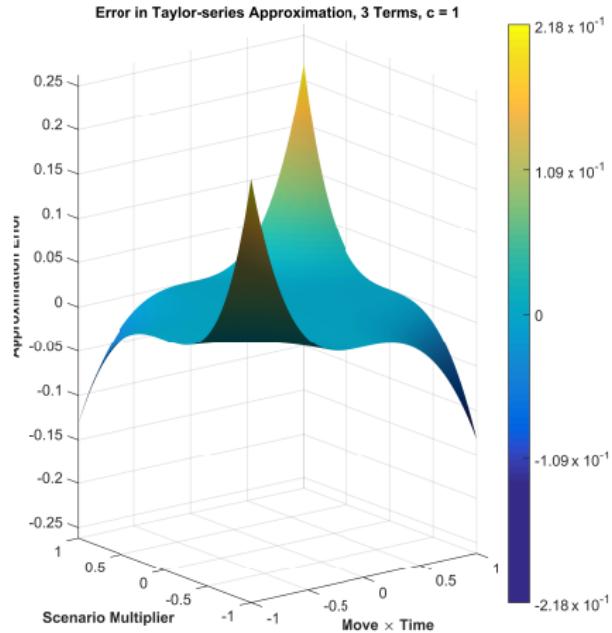
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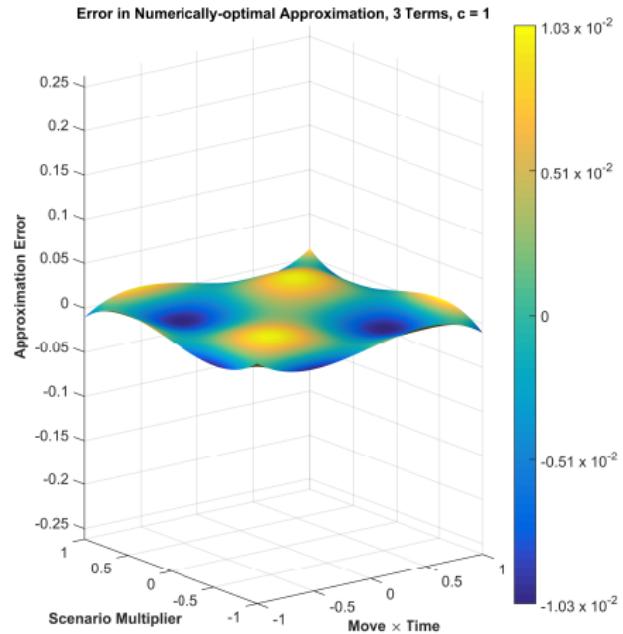
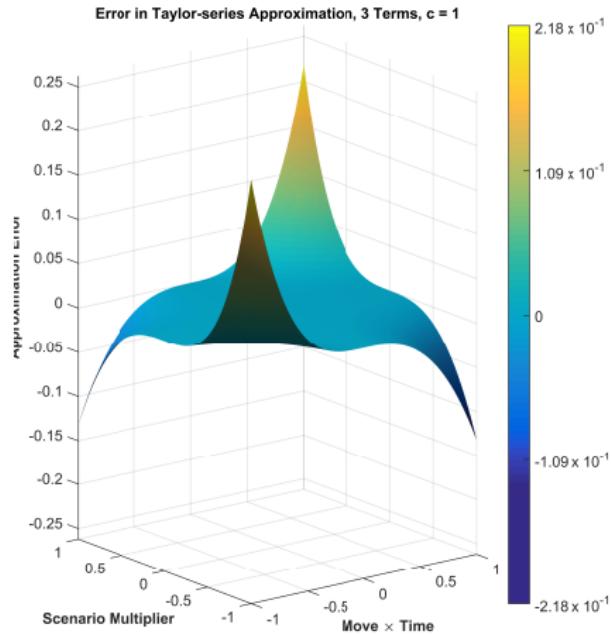
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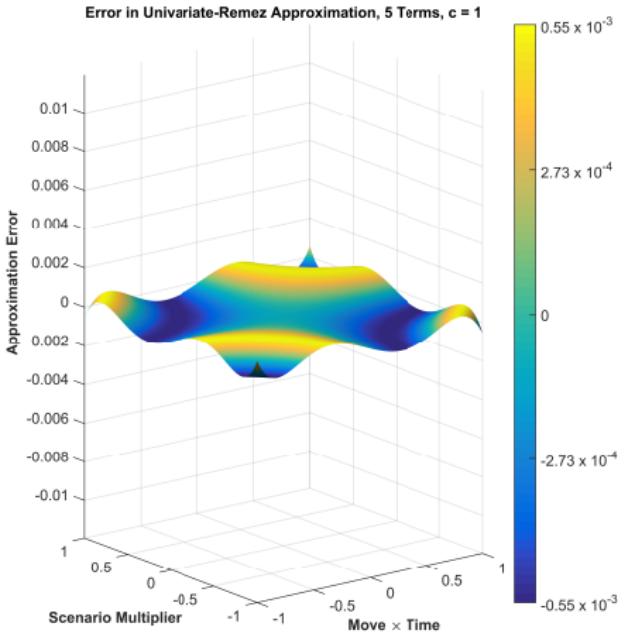
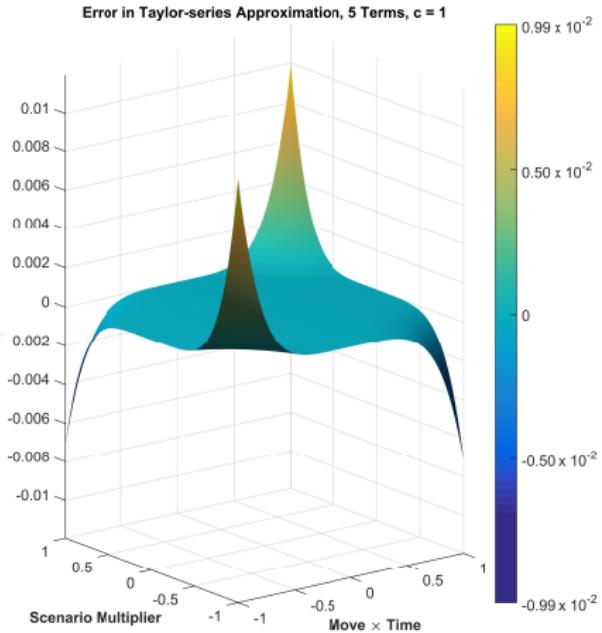
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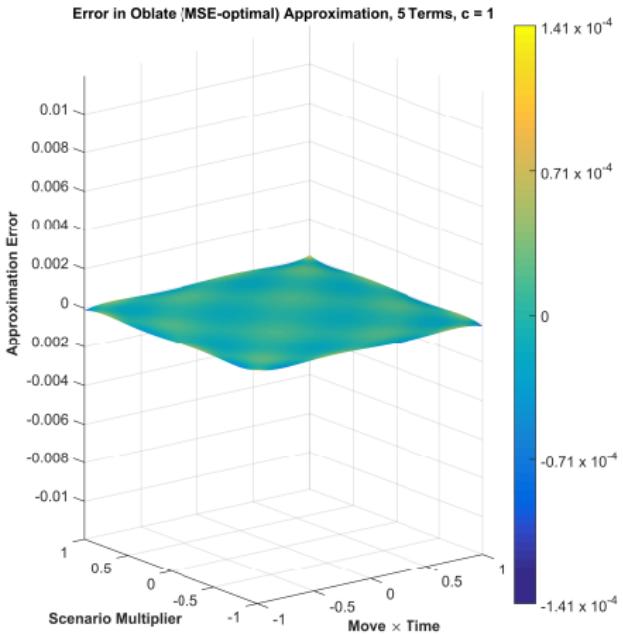
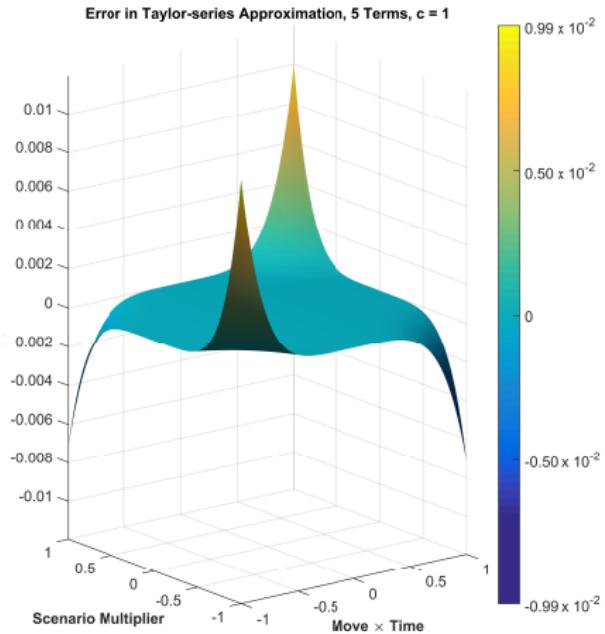
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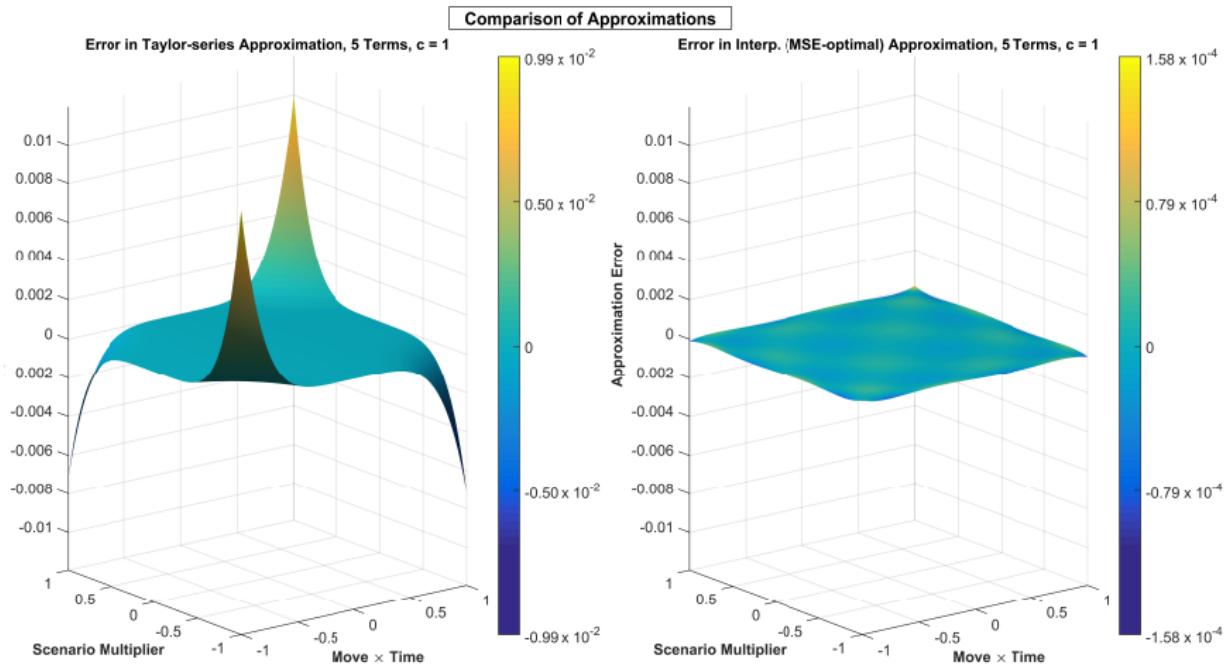
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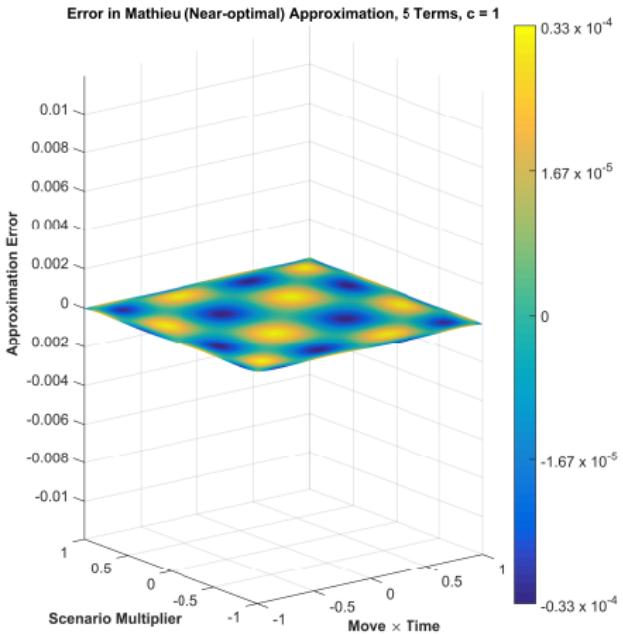
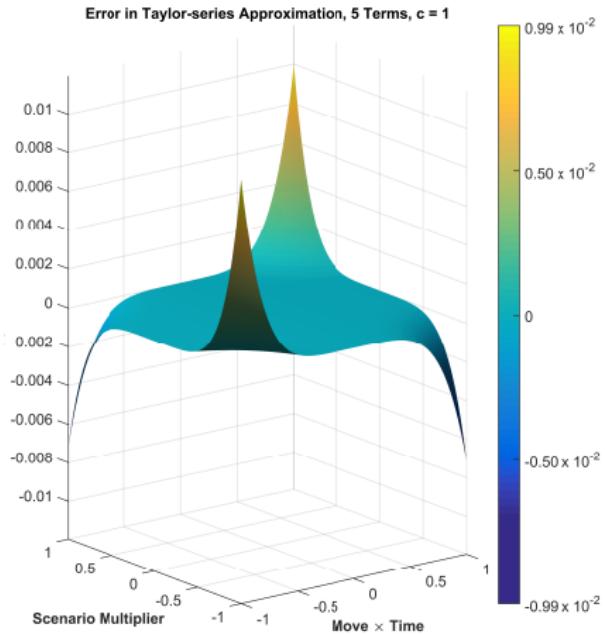
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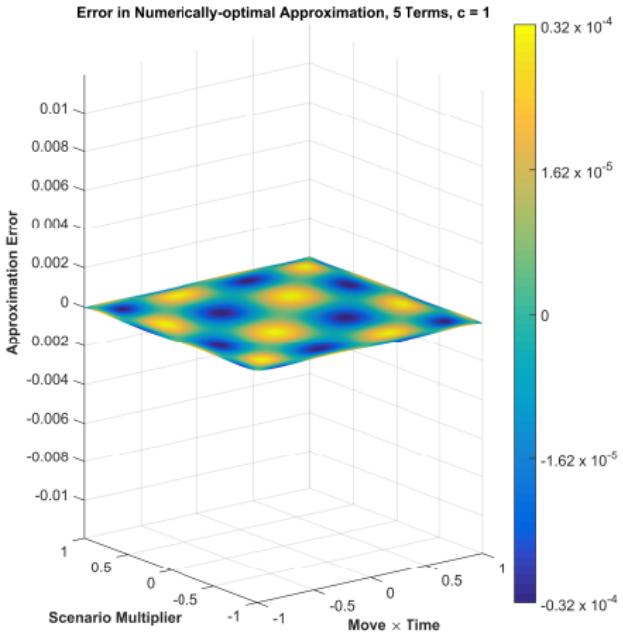
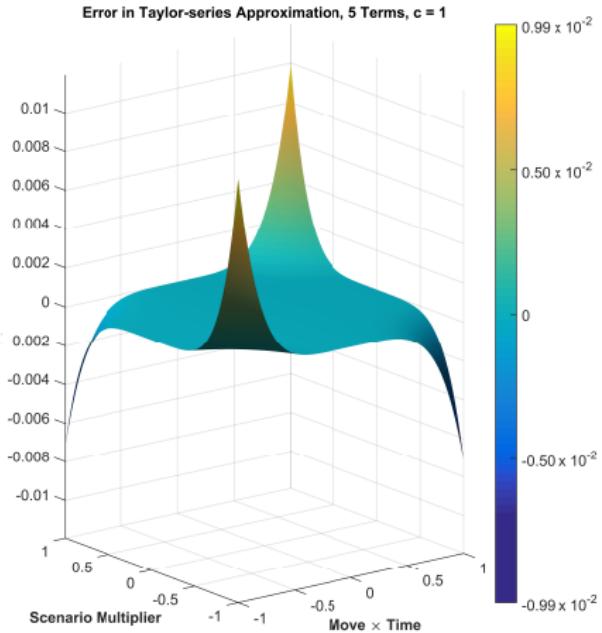
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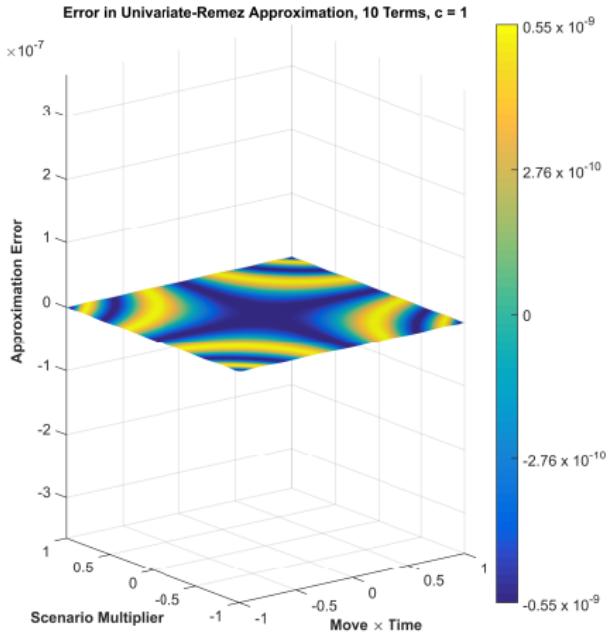
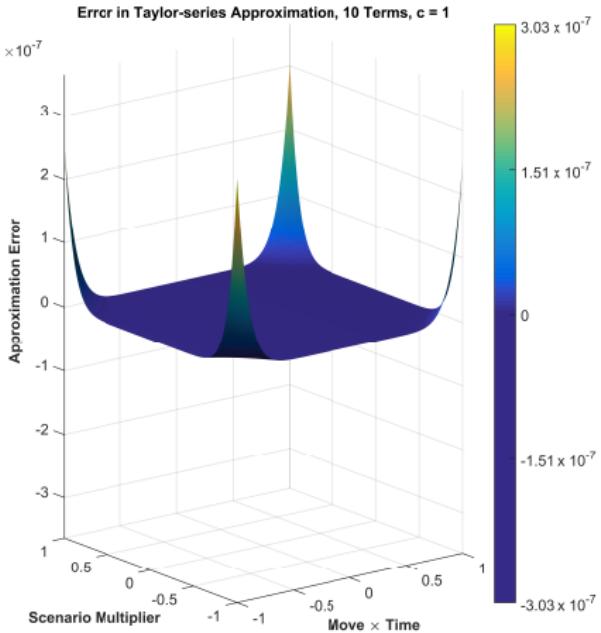
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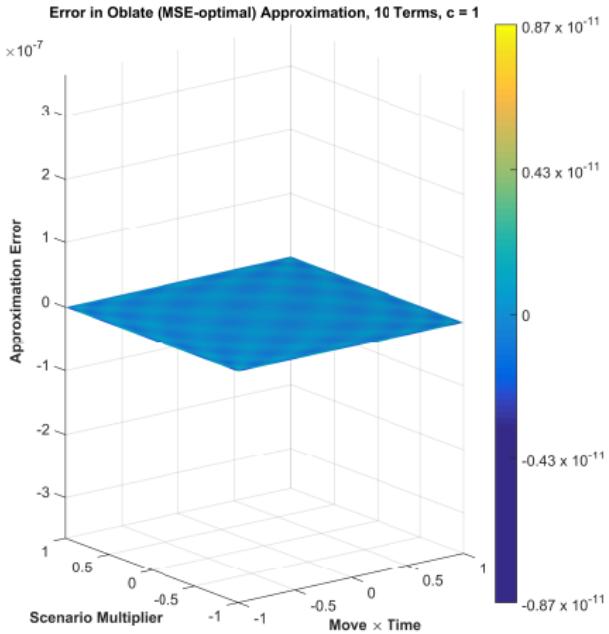
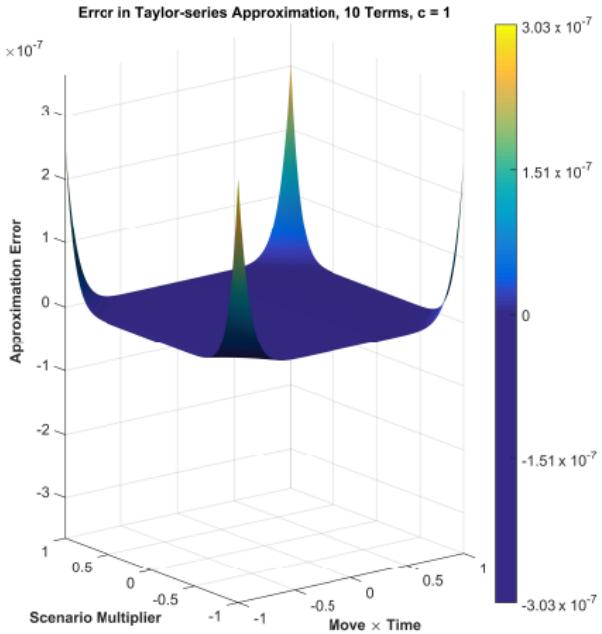
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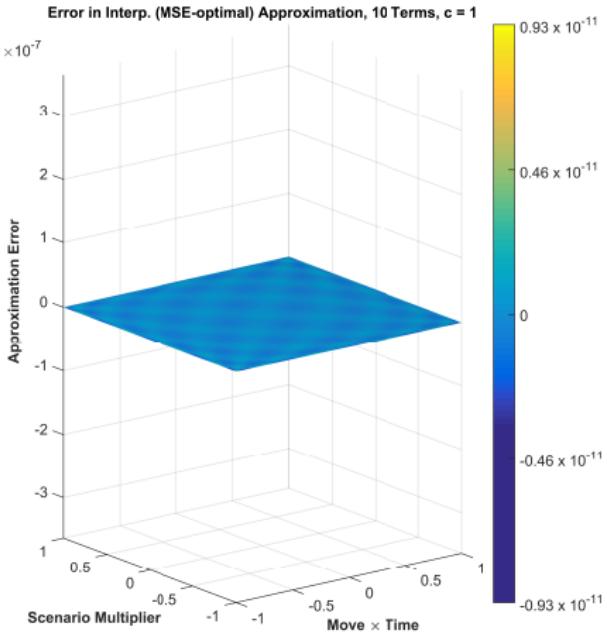
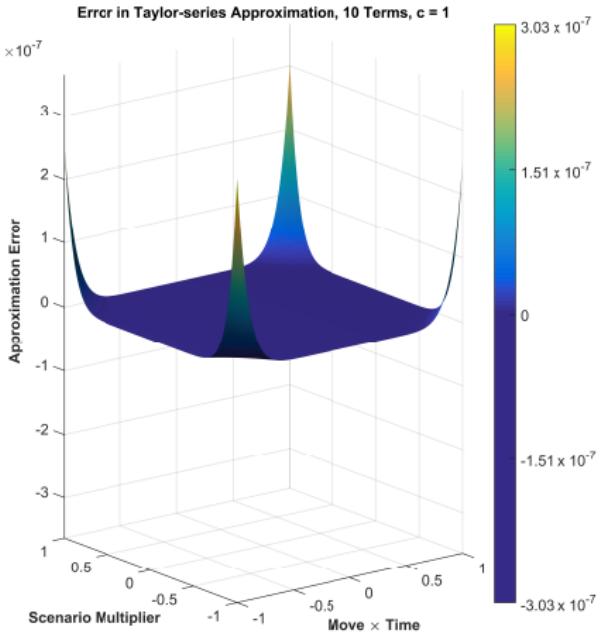
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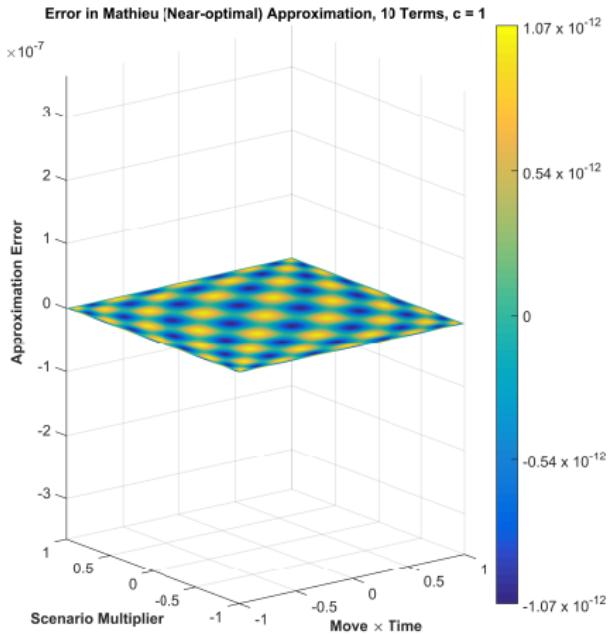
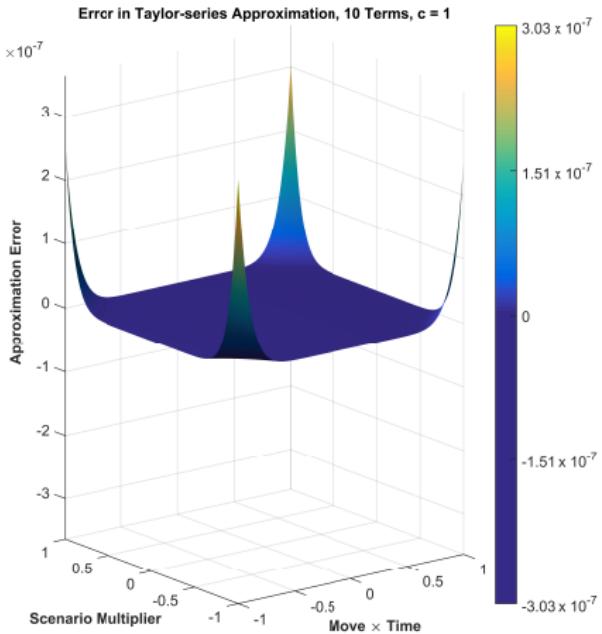
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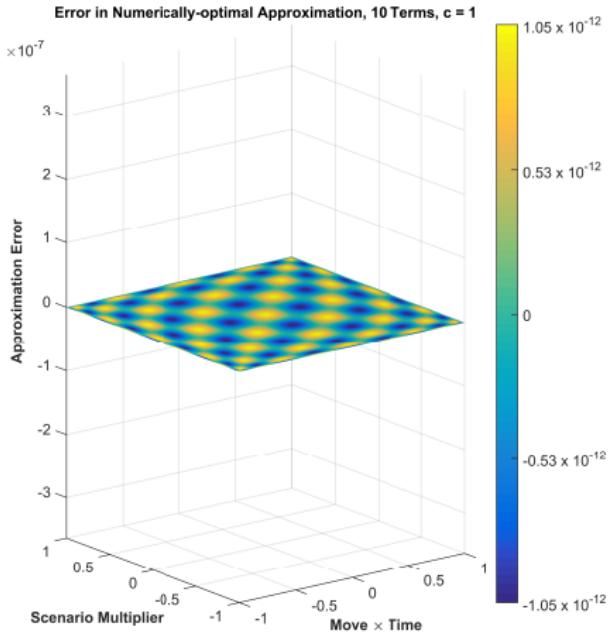
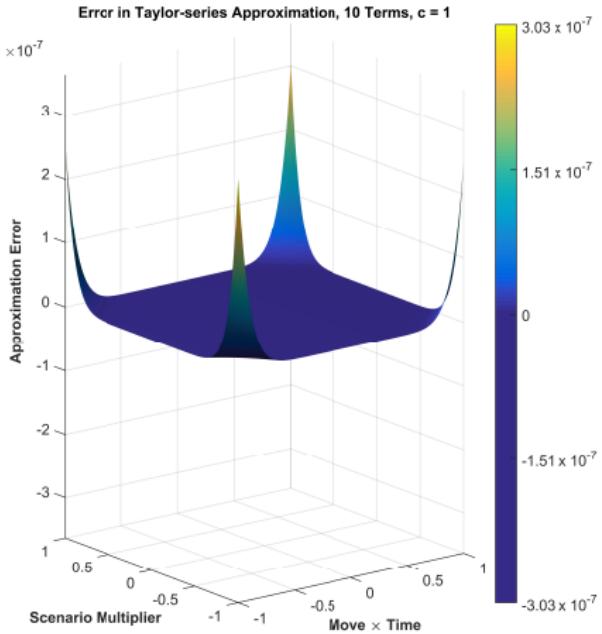
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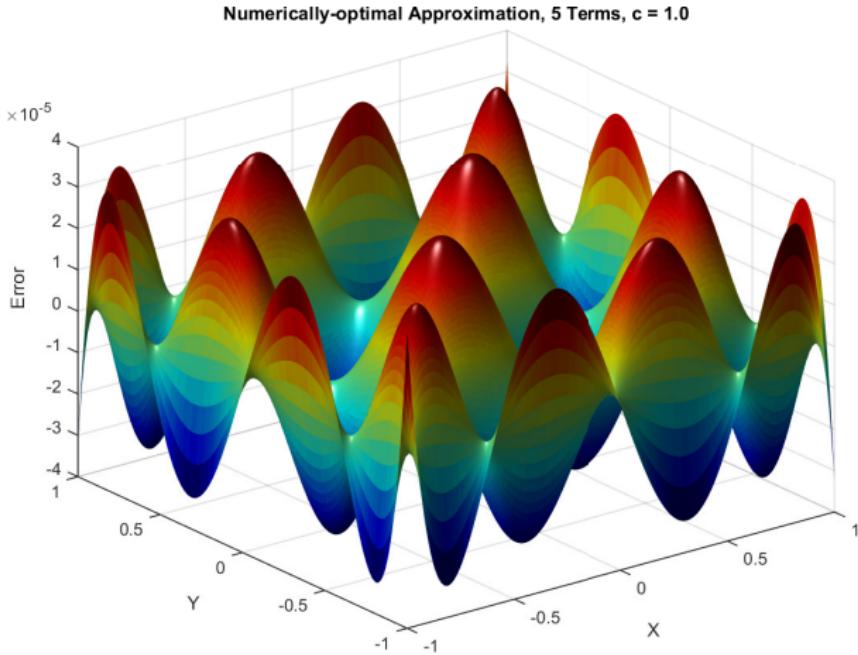


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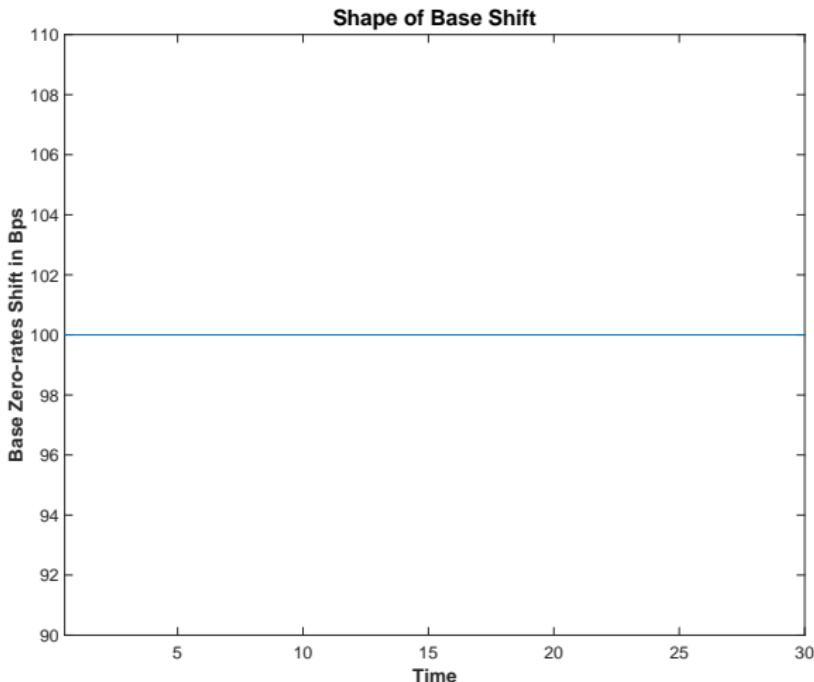


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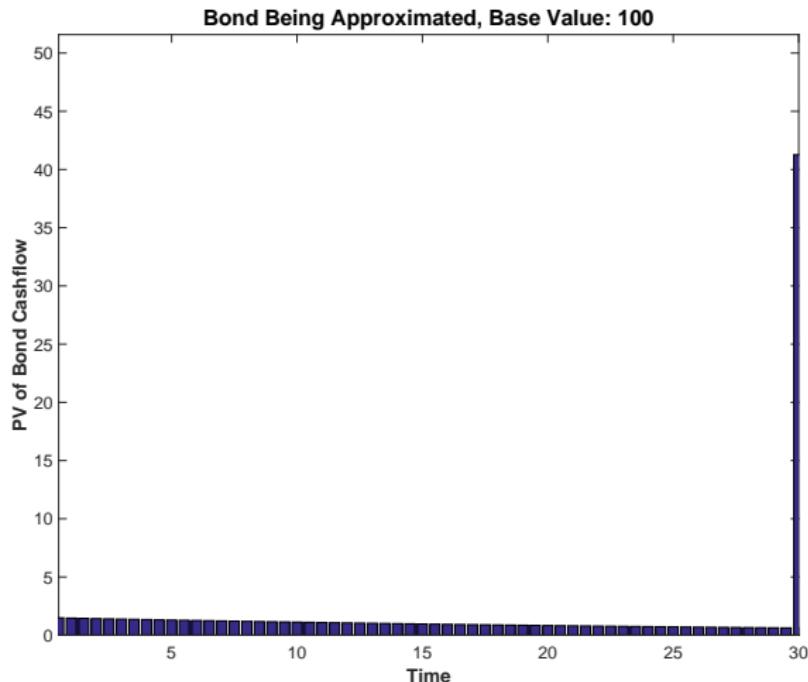


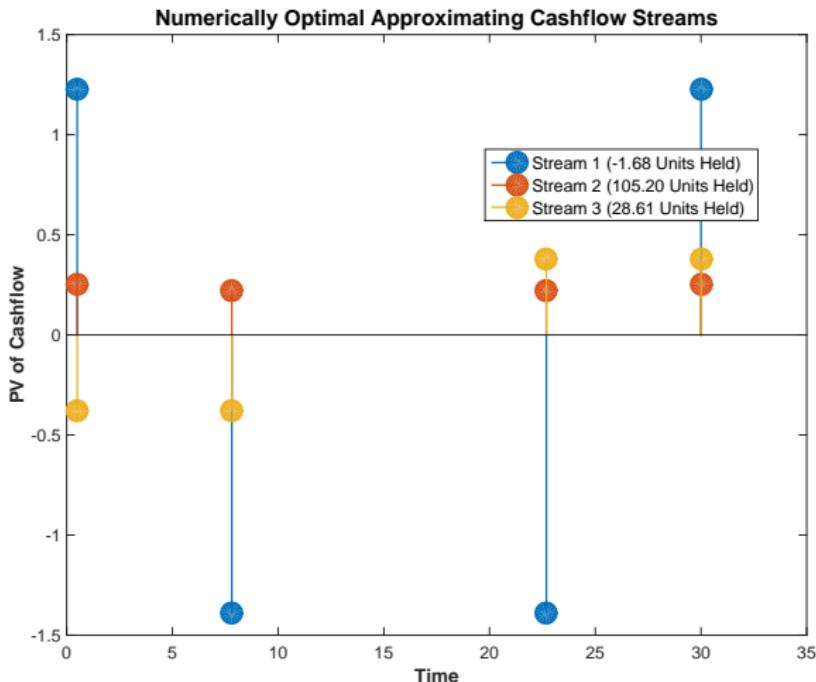
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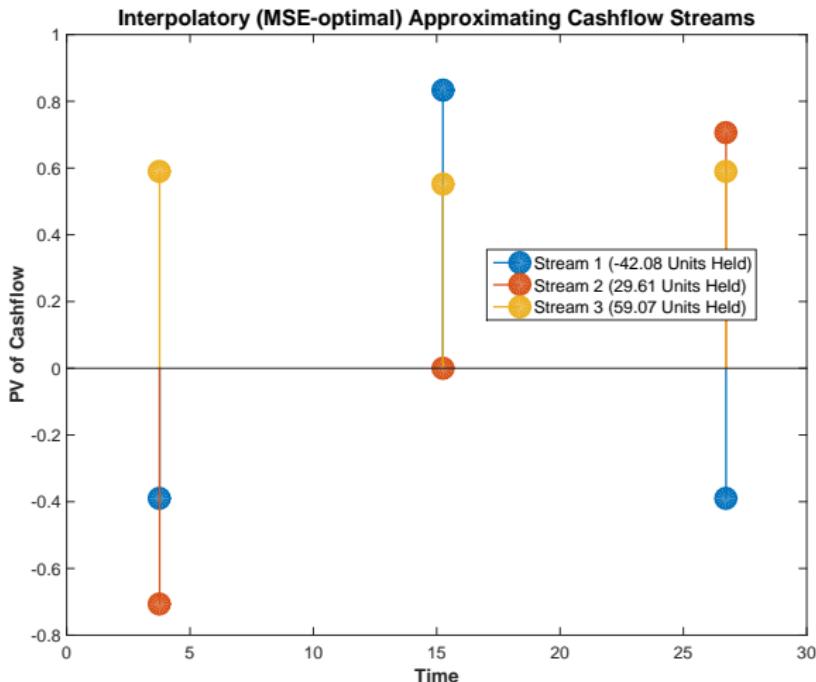
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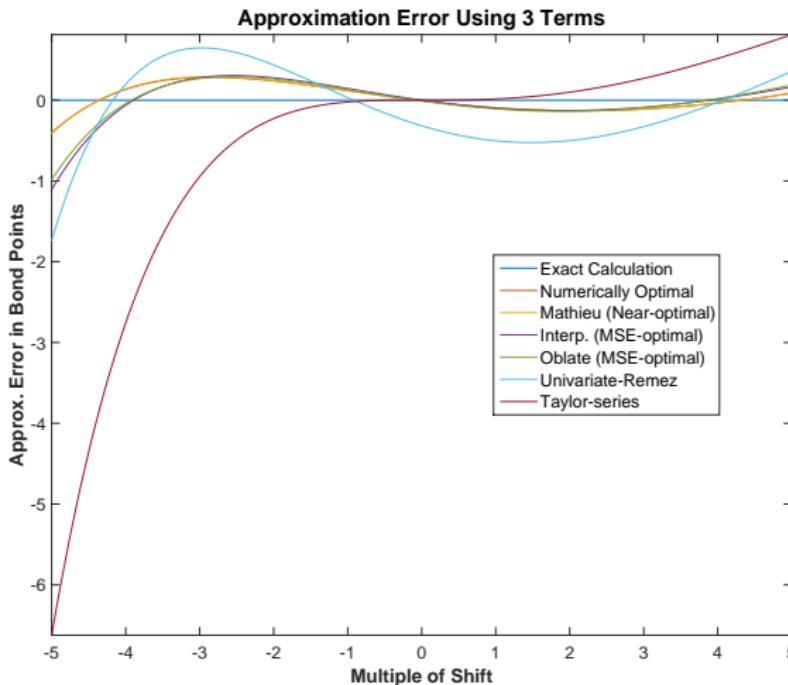


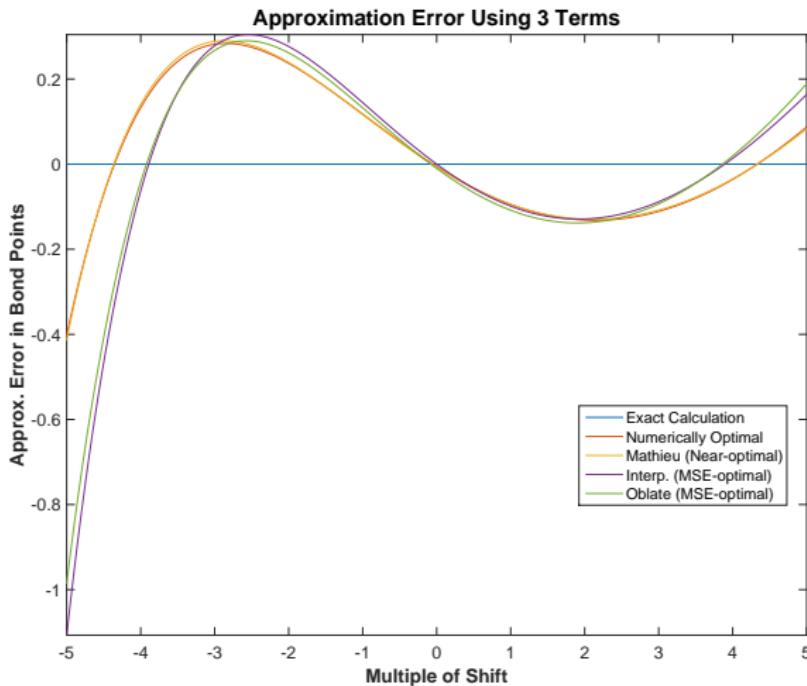
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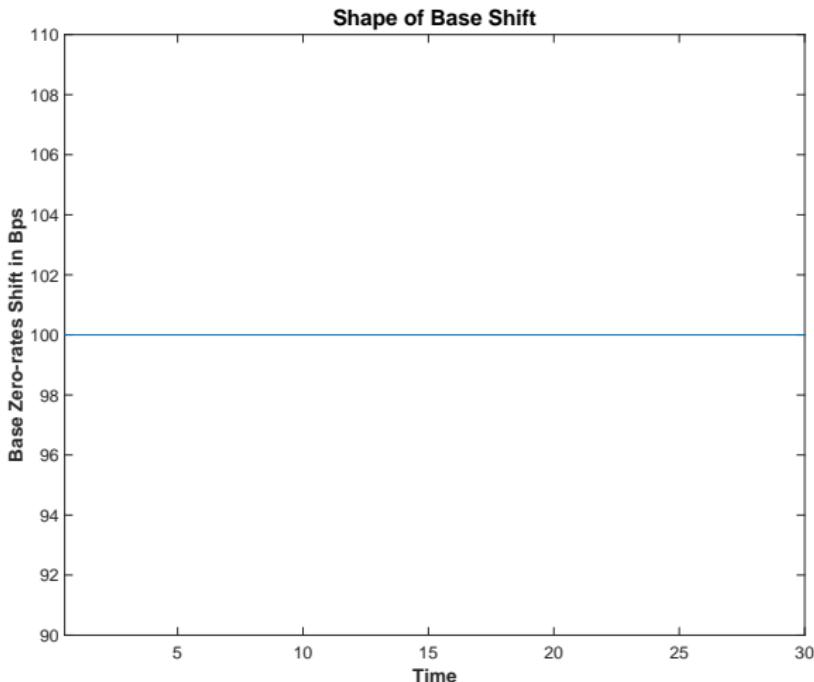


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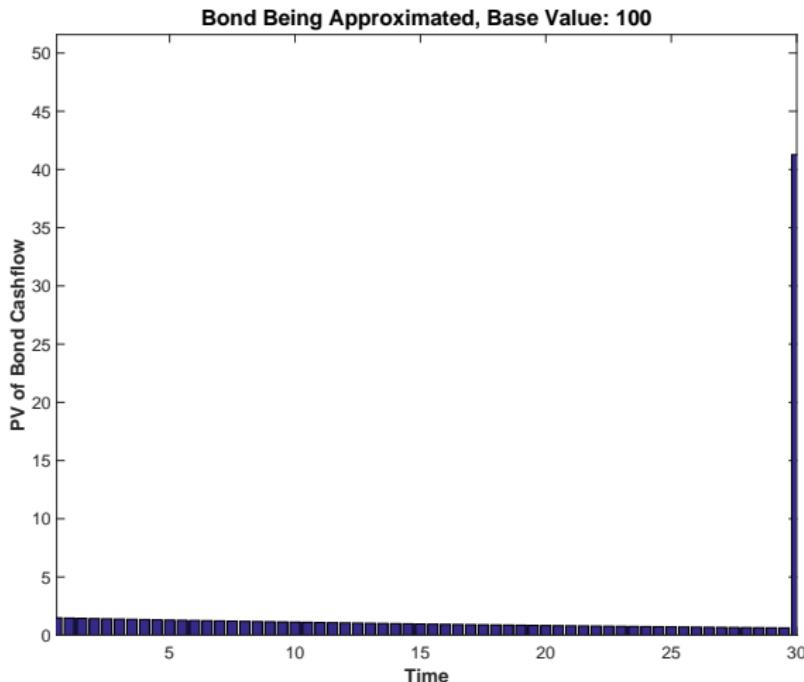


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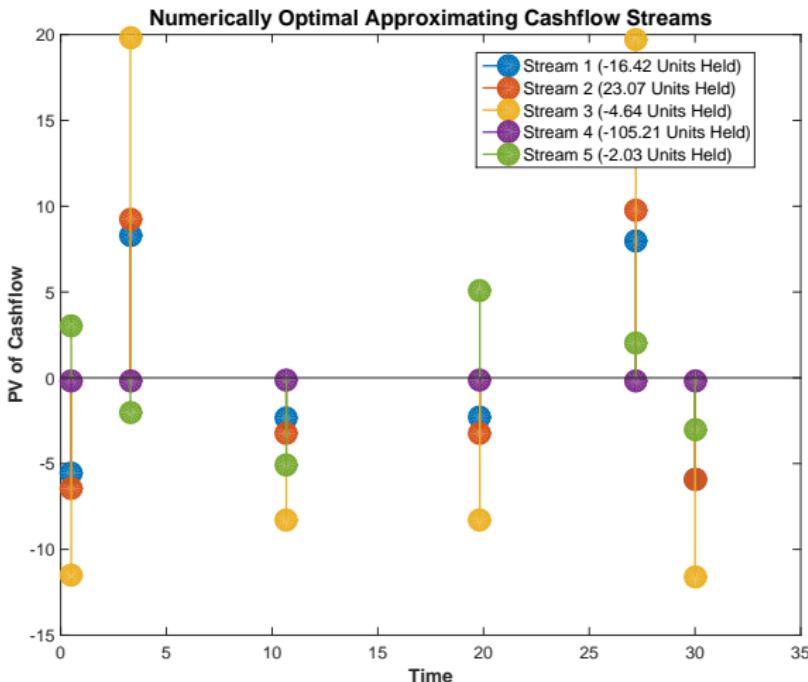


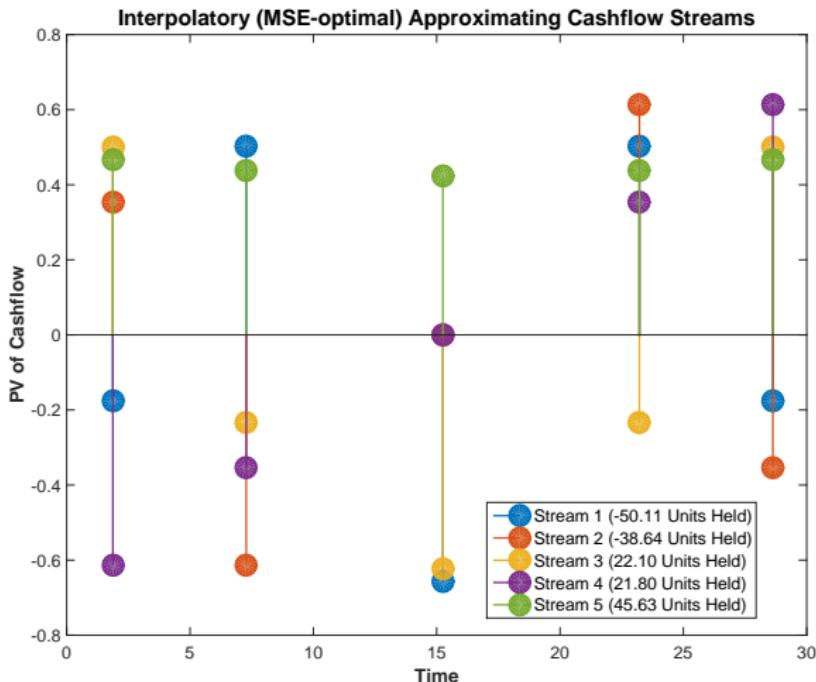
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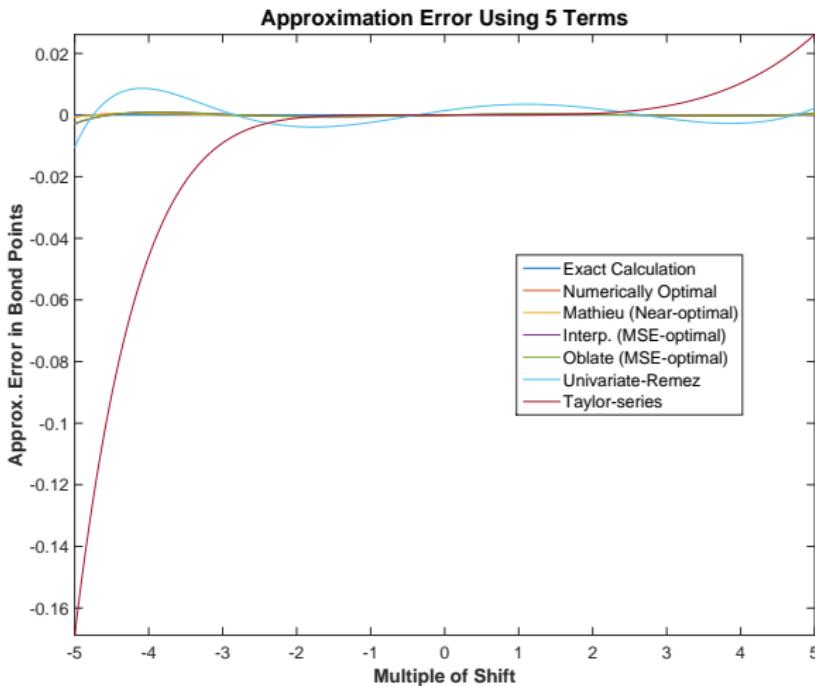


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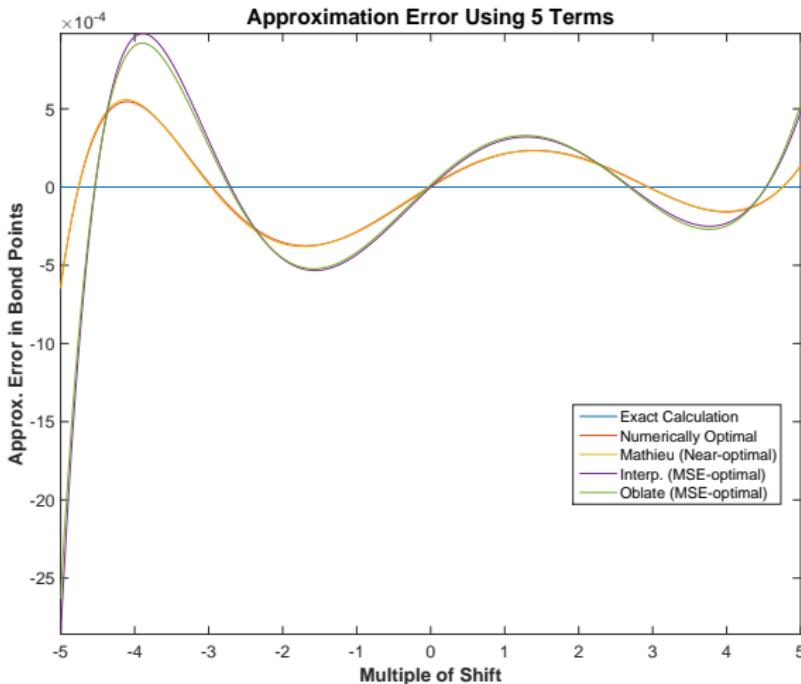


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n-Widths

- *n*-width theory measures how well *n*-dimensional sets can approximate infinite-dimensional sets (Kolmogorov (1936), Pinkus (1985))
- Prior to my work, applications were not financial: splines, signal processing, statistics
- I consider sets of scenario profiles generated by cashflow streams with gross present value of one
- Use the theory of *n*-widths to study optimal approximations to these sets

Total Positivity

- **Totally positive** functions have variation-diminishing properties when used to weight averages
- Applications in approximation problems, differential equations, statistics, and mechanism design
- Theory of ***n*-widths for totally positive kernels** provides important building blocks (reviewed in Pinkus (1985))
- Examples: $\exp(cx)$ (the exponential product kernel) on \mathbb{R}^2 , $\frac{1}{x+y}$ on $[0, \infty) \times (0, \infty)$, and $\exp(-(x-y)^2)$ on \mathbb{R}^2

Chebyshev Systems

- $n + 1$ functions form a **Chebyshev system** if any nontrivial linear combination of the functions has $\leq n$ zeros
- Must always be functions of one variable, effectively:
Mairhuber (1956)
- The approximation from a **Chebyshev system** that minimizes worst-case error is unique and its approximation error equioscillates
- Examples: polynomials, exponentials, trigonometric functions, eigenfunctions of some differential operators
- The **Remez method** works for these systems and shows how to compute best approximations

A Topological Result and A Lower Bound

- The **Borsuk Antipodality Theorem** (Borsuk (1933))
- Classic n -width result (Tikhomirov (1960)) uses this to lower-bound approximation errors
- I show that any list of $n + 1$ numbers $\rho_i \in [-1, 1]$ provides the lower bound on worst-case approximation error of

$$\min_{\mathbf{a}: \sum_{i=1}^{n+1} |a_i| = 1} \max_{y \in [-1, 1]} \left| \sum_{i=1}^{n+1} a_i \exp(c\rho_i y) \right| \quad (3)$$

for the exponential product operator $\exp(cx)$,
 $x, y \in [-1, 1]$.

Finding the Largest Lower Bound

- I develop a nontrivial variant of the Remez method to solve (3)
- Then I use this variant as a building block in a fixed-point procedure that, empirically, finds the points ρ that give the largest lower bound of this type

Sample Output: Fixed Point Procedure I

```
» [wRhoVector, wCoeffs, wAmatrix, wNodes, wErrorsAtNodes,
   wDiscrepancy, wNumIter] = worstRhoForExp(3, 1 , 1e-15 , 100
   , true)
wRhoVector =
-1 0 1
wCoeffs =
-0.989586650561472 0.0000000000000001
0.143938393177520
wAmatrix =
0.596243191512492 -1.669693283914562
0.596243191512493 -0.461226118145786
0.000000000000000 0.461226118145786
```

Sample Output: Fixed Point Procedure II

-0.323775931284113 -0.231239110157009

-0.323775931284113

wNodes =

-1.0000000000000000 0.0000000000000000

1.0000000000000000

wErrorsAtNodes =

-0.345745818123157 0.345745818123157

-0.345745818123156

wDiscrepancy =

0

wNumIter =

1

Sample Output: Computing the Lower Bound I

```
» [ bCoeffs , bNodes , bErrorsAtNodes , bDiscrepancy ,
bNumIter] = borsukLowerBound(wRhoVector , 1 , 1e-15 , 100)
bCoeffs =
-0.220114957312587 0.559770085374825
-0.220114957312587
bNodes =
-1 0 1
bErrorsAtNodes =
-0.119540170749650 0.119540170749650
-0.119540170749650
bDiscrepancy =
0
```

Sample Output: Computing the Lower Bound II

bNumIter =

0

Example of Numerical Results for the Lower Bound

	c	n	Lower Bound
	0.900	1	1.026516725708175
	0.900	2	0.097694100180484
	0.900	3	0.007500615564943
	0.900	4	0.000424290711434
	0.900	5	0.000019140824085
	0.900	6	0.000000718728548
	0.900	7	0.000000023120075
	0.900	8	0.000000000650579
	0.900	9	0.000000000016270
	0.900	10	0.000000000000366
	0.900	11	0.000000000000007
	0.900	12	0.000000000000000

Numerically-optimal Approximation: Motivation

- Speculate: the optimal rank- n linear approximation may also be optimal in approximating a rank- $(n + 1)$ operator
- **Imagine a game:** market plays a grid with $n + 1$ values of x and $n + 1$ values of y
- What is the best rank- n response?
- Hope that the Bernstein, Gel'fand, Kolmogorov, and linear n -widths are identical
- Apply the results of Micchelli & Pinkus (1979)
- Upper bound on approximation error using interval analysis (**computationally intensive**)

Numerically-optimal Approximation: Construction

To construct $P_{c,n}^*(x, y)$ I find very particular:

- ρ^\dagger , a $(n + 1) \times 1$ vector (from fixed-point procedure)
- a $(n + 1) \times n$ matrix V^*

Then

$$P_{c,n}^*(x, y) = \exp(cx\rho^\dagger)^T V^* V^{*T} \exp(c\rho^\dagger y) \quad (4)$$

$$= \sum_{i=1}^n f_i(x) f_i(y), \quad (5)$$

where $f_i(x) = \sum_{j=1}^{n+1} V_{ji}^* \exp(cx\rho_j^\dagger)$.

Numerically-optimal Approximation: Interpretation

- The numerically-optimal n -dim'l. approximating space is spanned by n simple financial instruments
- Each of these instruments is a **cashflow stream**
- All n of these streams have their cashflows on a common set of $n + 1$ times
- The approximating portfolio is **linear** in the cashflow stream

Results of Interval Analysis

	c	n	Lower Bound	Numerically-optimal
	0.900	1	1.026516725708175	1.026516725708176
	0.900	2	0.097694100180484	0.097694100180486
	0.900	3	0.007500615564943	0.007500615564952
	0.900	4	0.000424290711434	0.000424290711437
	0.900	5	0.000019140824085	0.000019140824093
	0.900	6	0.000000718728548	0.000000718728562
	0.900	7	0.000000023120075	0.000000023120094
	0.900	8	0.000000000650579	0.000000000650595
	0.900	9	0.000000000016270	0.000000000016291
	0.900	10	0.000000000000366	0.000000000000377
	0.900	11	0.000000000000007	0.000000000000012
	0.900	12	0.000000000000000	0.000000000000007

Interpolatory Approximation: Motivation

- Optimal approximations are generally not unique
- Eigenfunction expansions can be truncated to obtain optimal approximations for average squared error
- The eigenfunctions of P_c are the **oblates** (oblate spheroidal wave functions of order zero)
- Melkman & Micchelli (1978) showed that totally positive kernels also have **interpolatory** approximations that are optimal for average squared error
- Interpolate at the n **zeros** of the $(n + 1)^{\text{st}}$ eigenfunction

Interpolatory Approximation: Construction

$$P_{c,n}^{\text{Interp}}(x, y) = \begin{pmatrix} \exp(cx\eta_1) \\ \vdots \\ \exp(cx\eta_n) \end{pmatrix}^T U^T \Lambda^{-1} U \begin{pmatrix} \exp(c\eta_1 y) \\ \vdots \\ \exp(c\eta_n y) \end{pmatrix}$$

where the η_i are the zeros of the $(n+1)^{\text{st}}$ oblate (it has index n , since indexing starts with zero for the oblates) and

$$\begin{pmatrix} \exp(c\eta_1\eta_1) & \cdots & \exp(c\eta_1\eta_n) \\ \vdots & \ddots & \vdots \\ \exp(c\eta_n\eta_1) & \cdots & \exp(c\eta_n\eta_n) \end{pmatrix} = U^T \Lambda U$$

Interpolatory Approximation: Interpretation

- The rank- n interpolatory approximation uses a portfolio of a fixed set of n approximating cashflow streams
- All n approximating cashflow streams have cashflows at the same n times
- The approximating portfolio is linear in the target cashflow stream
- Different cashflow streams have distinct approximating holdings of the same n approximating cashflow streams

Mathieu Approximation: Motivation

- In one-variable problems, there is a **weighting** that gives excellent polynomial approximations (in worst-case error)
- These are Chebyshev polynomial approximations (Trefethen (2013))
- Will a similar result hold for my two-variable problem?
- I show and exploit: Mathieu differential operator **commutes** with exponential cosine product operator

The Mathieu Operator

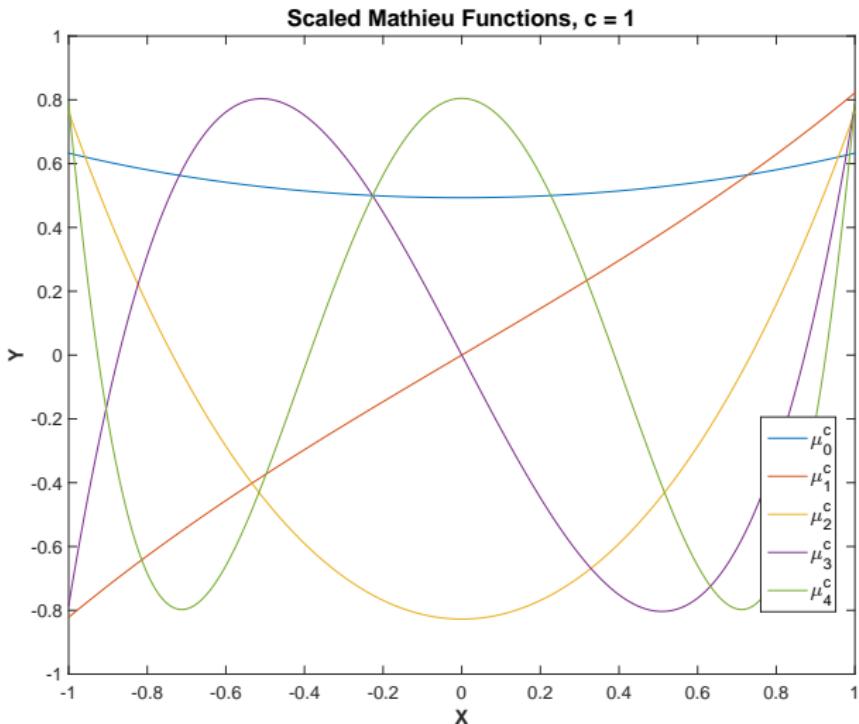
$$M_c [f] (t) \equiv -\frac{d^2 f}{dt^2} (t) - \frac{c^2}{2} \cos(2t) f(t), \quad (6)$$

though its standard form is written in terms of $q = -\frac{c^2}{4}$.

- The eigenfunctions of this differential operator are the **Mathieu functions**
- They are special functions of classical mathematical physics

Mathieu Approximation: Construction

- Expand Mathieu functions in series of Chebyshev polynomials (minor variation on traditional Fourier series)
- Result is two symmetric tridiagonal eigenproblems for series coefficients (eigenvectors)
- Solve these by truncation (as in Zhang & Jin (1996))
- Use eigenfunction identity to get the eigenvalues for the weighted exponential product operator (new)



Oblate Approximation: Motivation

- Eigenfunction expansions can be truncated to give average-squared-error optimal approximations
- What are the eigenfunctions of the exponential product operators?
- Use a commuting differential operator, the oblate spheroidal wave operator of order zero (Zayed (2007) notes this briefly)
- The eigenfunctions are the oblate spheroidal wave functions of order zero (**oblates**)

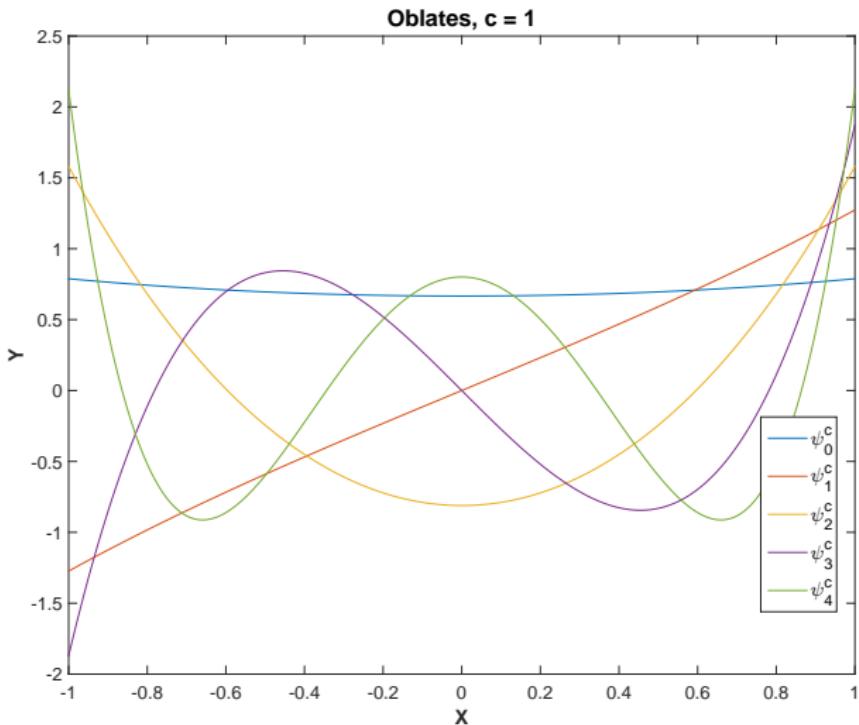
The Oblate Spheroidal Wave Operator

$$L_c [f] (x) \equiv -\frac{d}{dx} \left(\left(1 - x^2\right) \frac{df}{dx} (x) \right) - c^2 x^2 f(x) \quad (7)$$

- The eigenfunctions of this differential operator are the **oblates** (full name: oblate spheroidal wave functions of order zero)
- They are special functions of classical mathematical physics

Oblate Approximation: Construction

- Expand the oblates in series of Legendre polynomials (standard approach)
- Result is two symmetric tridiagonal eigenproblems for series coefficients (eigenvectors)
- Solve these by truncation (as in Hodge (1970))
- Use eigenfunction identity to get the eigenvalues for the exponential product operator (as in prolate case, though new here)

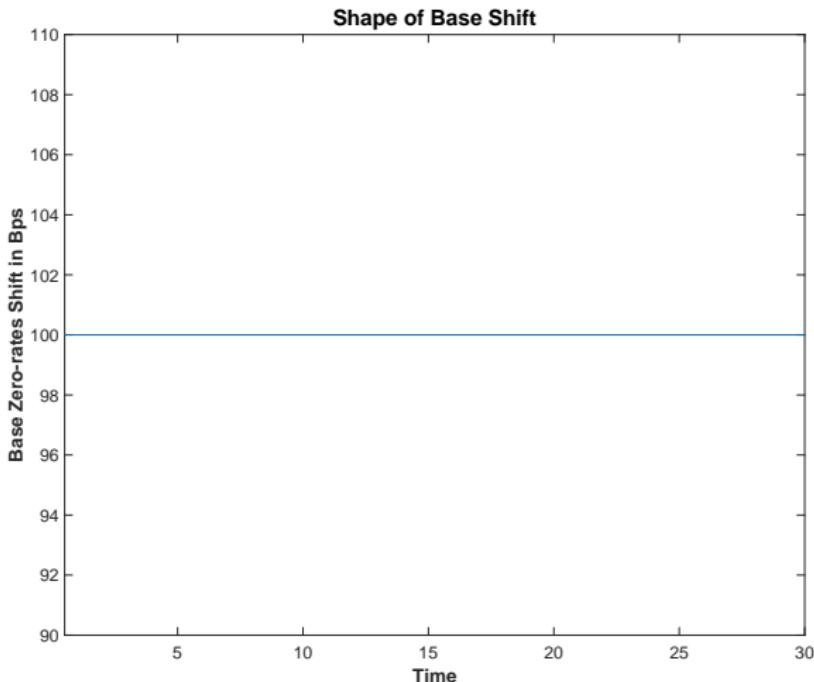


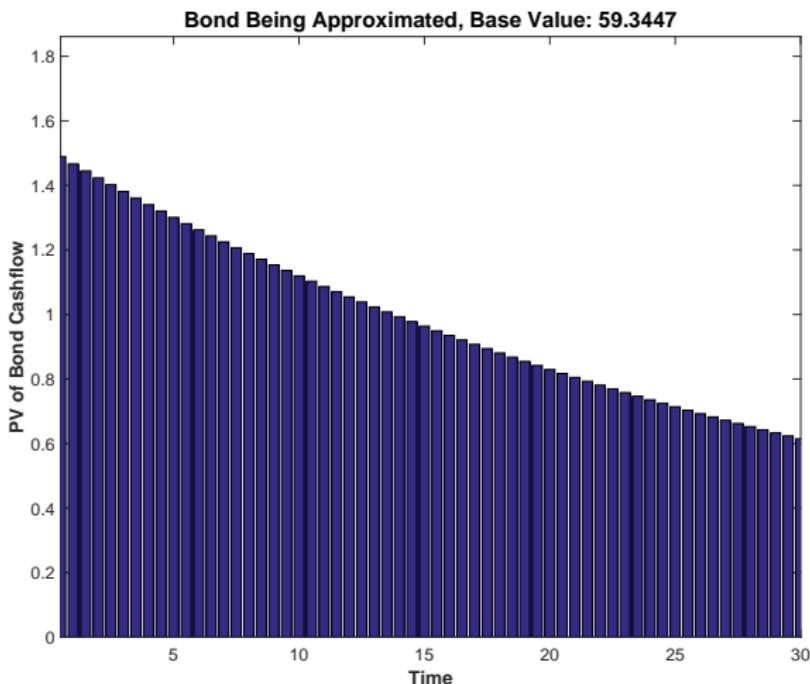
Comparisons of Approximations: Worst-case Error

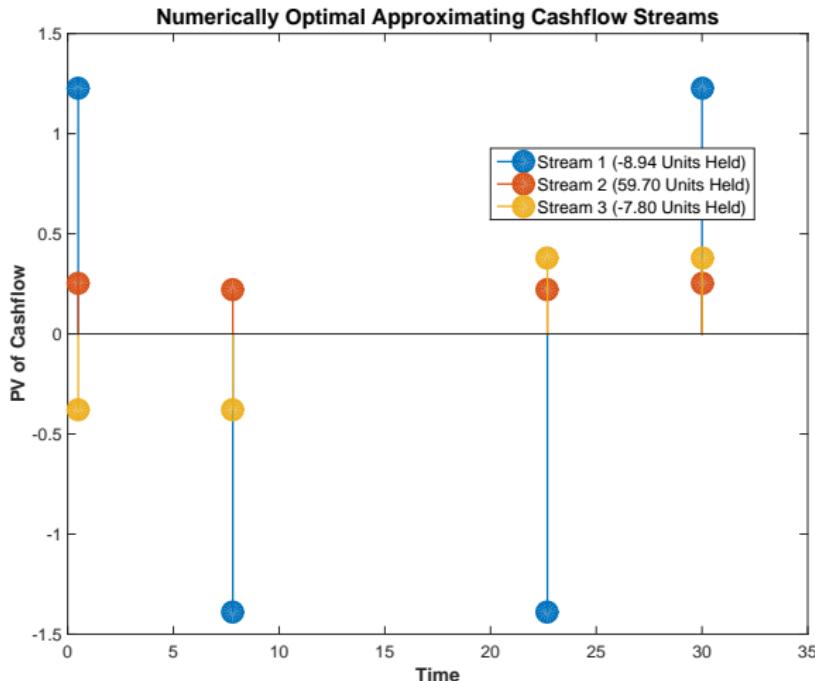
c	n	Num.-optimal	Mathieu	Taylor-series
0.900	1	1.026516725708176	1.142627510037619	1.459603111156950
0.900	2	0.097694100180486	0.123445382124493	0.559603111156950
0.900	3	0.007500615564952	0.008533145878606	0.154603111156950
0.900	4	0.000424290711437	0.000468349741963	0.033103111156950
0.900	5	0.000019140824093	0.000020725649064	0.005765611156950
0.900	6	0.000000718728562	0.000000768395225	0.000844861156950
0.900	7	0.000000023120094	0.000000024492435	0.000106748656950
0.900	8	0.000000000650595	0.000000000684441	0.000011848478378
0.900	9	0.0000000000016291	0.0000000000017024	0.000001172208289
0.900	10	0.000000000000377	0.000000000000381	0.000000104581280
0.900	11	0.000000000000012	0.000000000000008	0.000000008494849
0.900	12	0.000000000000007	0.000000000000000	0.0000000000633232

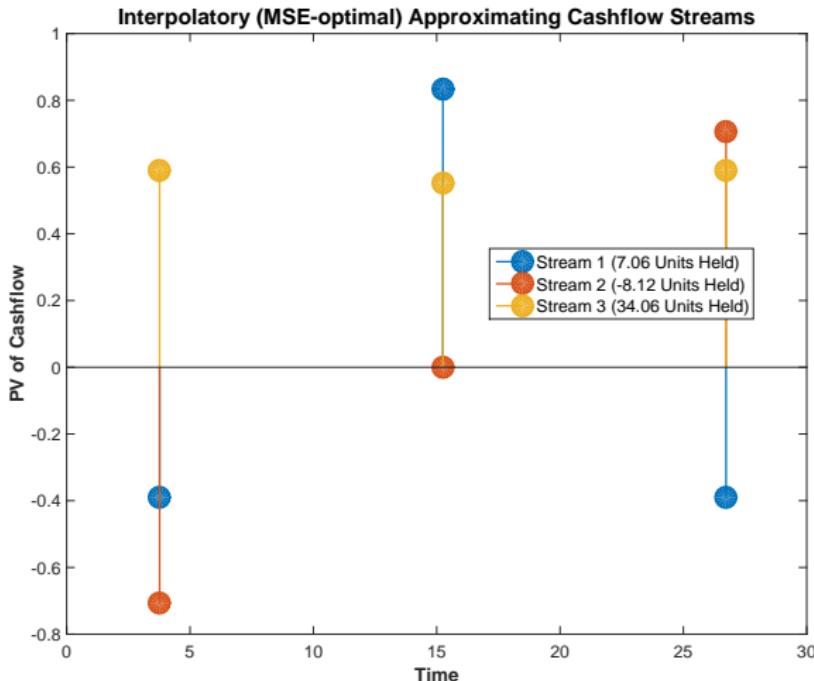
Comparisons of Approximations: RMSE

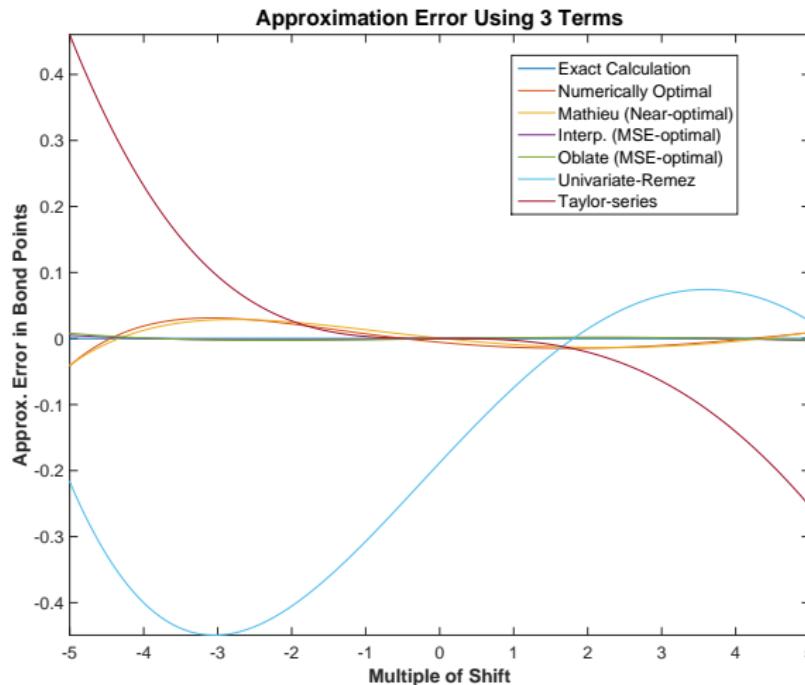
c	n	Interpolatory	Oblate	Taylor-series
0.900	1	0.630046304453232	0.630046304453232	0.630046304453232
0.900	2	0.072215718962917	0.072215718962917	0.167688553451906
0.900	3	0.005560981010366	0.005560981010366	0.035576959630715
0.900	4	0.000317559230476	0.000317559230476	0.006186015811949
0.900	5	0.000014430879283	0.000014430879283	0.000907144882218
0.900	6	0.000000544878914	0.000000544878914	0.000114798404740
0.900	7	0.000000017602704	0.000000017602704	0.000012764853380
0.900	8	0.000000000497000	0.000000000497000	0.000001265127833
0.900	9	0.0000000000012463	0.0000000000012463	0.000000113062188
0.900	10	0.000000000000281	0.000000000000281	0.000000009198080
0.900	11	0.000000000000006	0.000000000000006	0.0000000000686633
0.900	12	0.000000000000002	0.000000000000001	0.0000000000047350

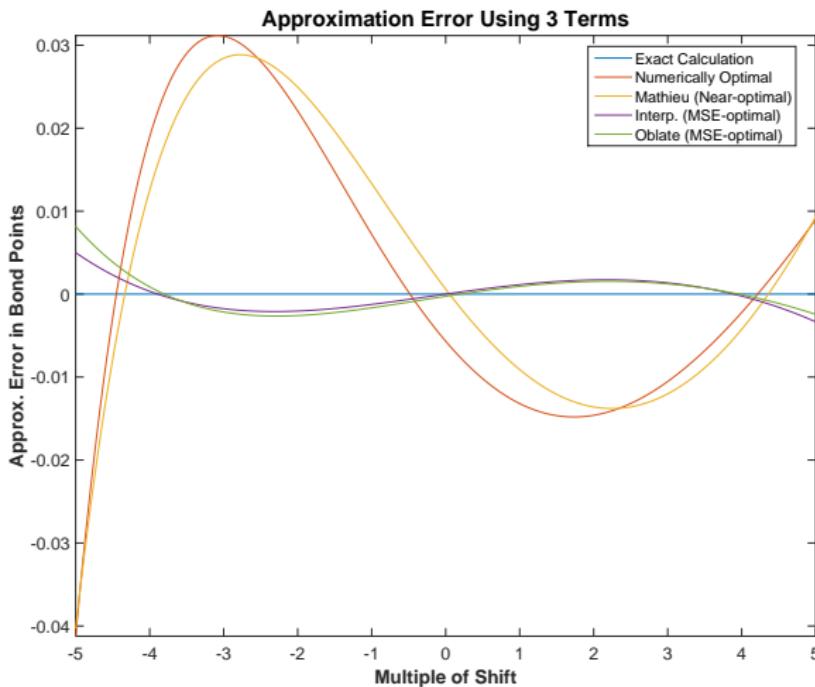


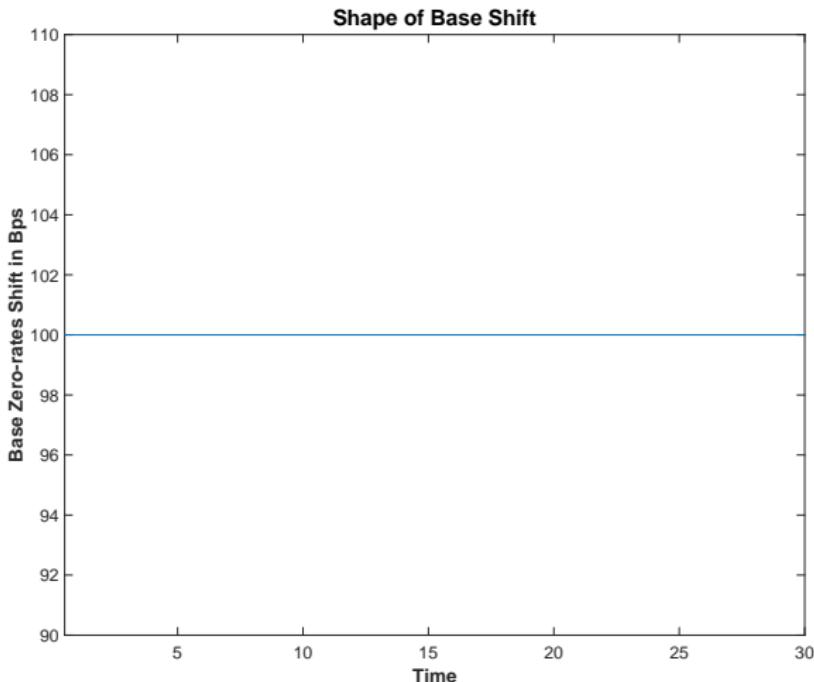


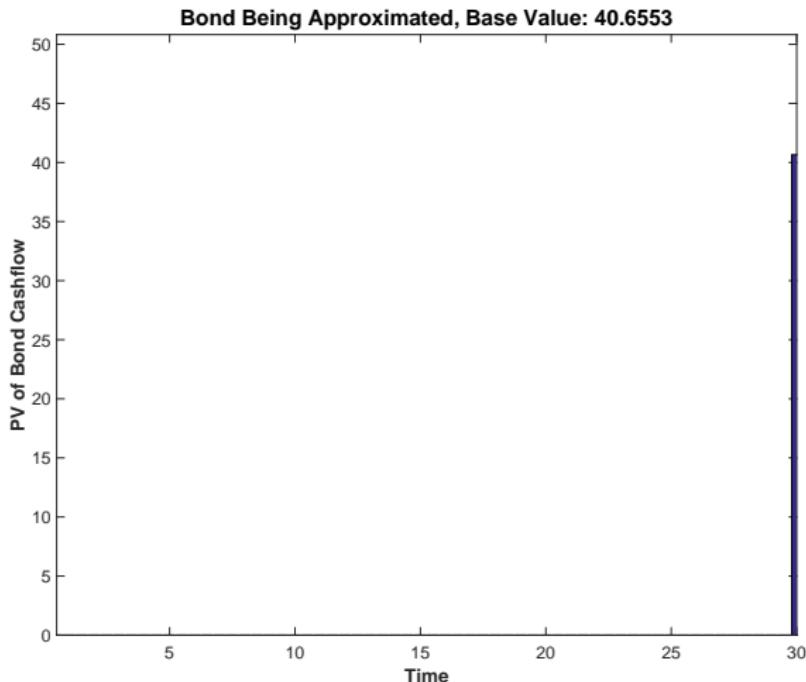


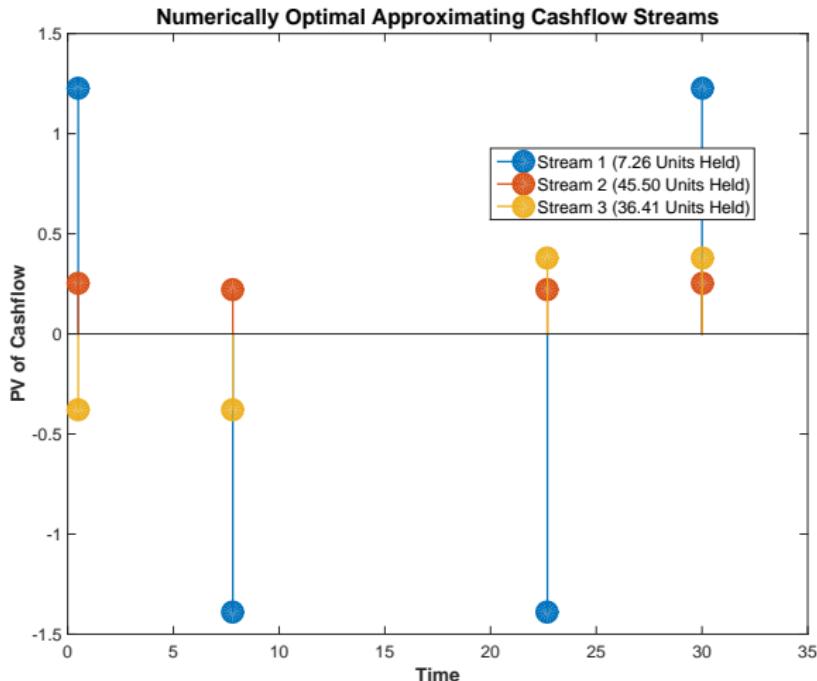


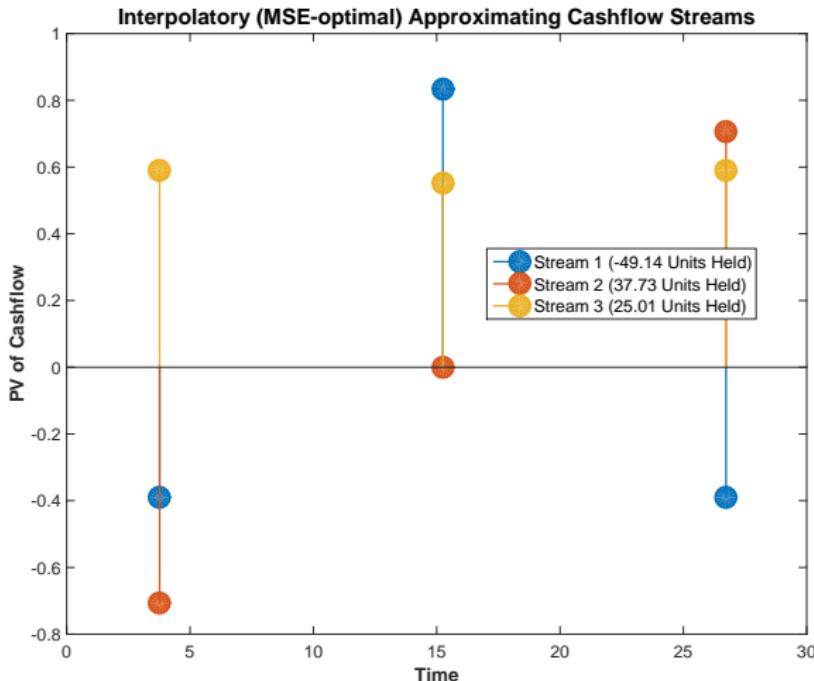


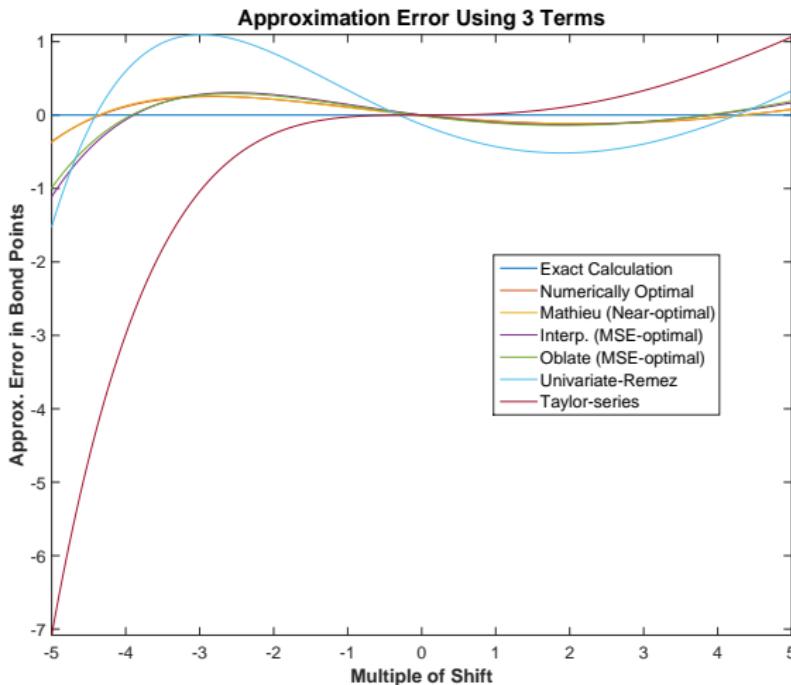


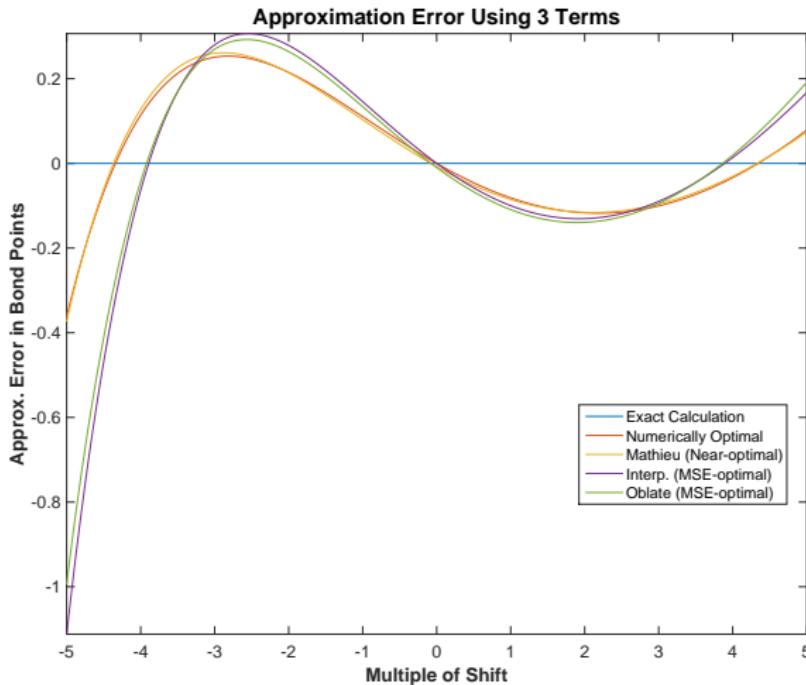


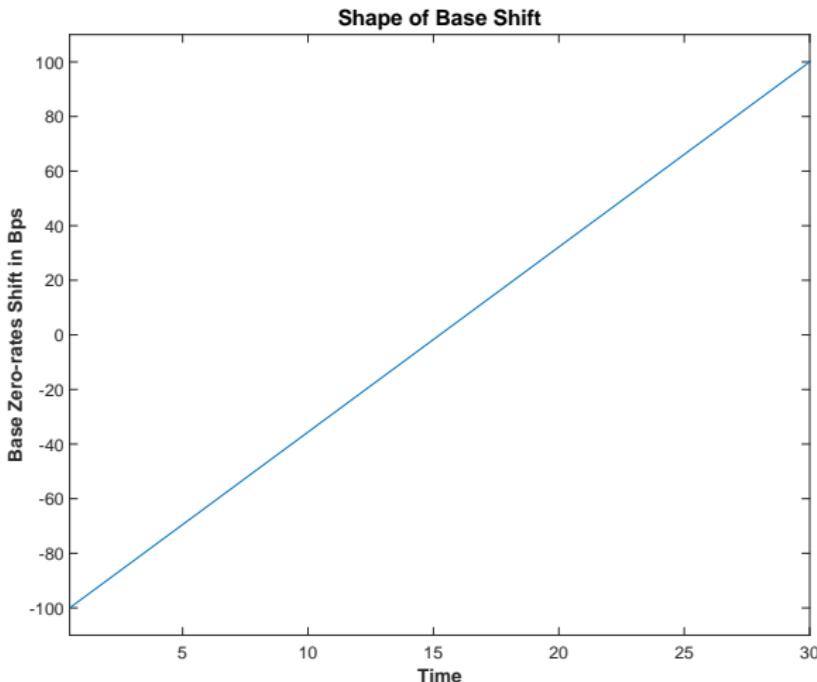


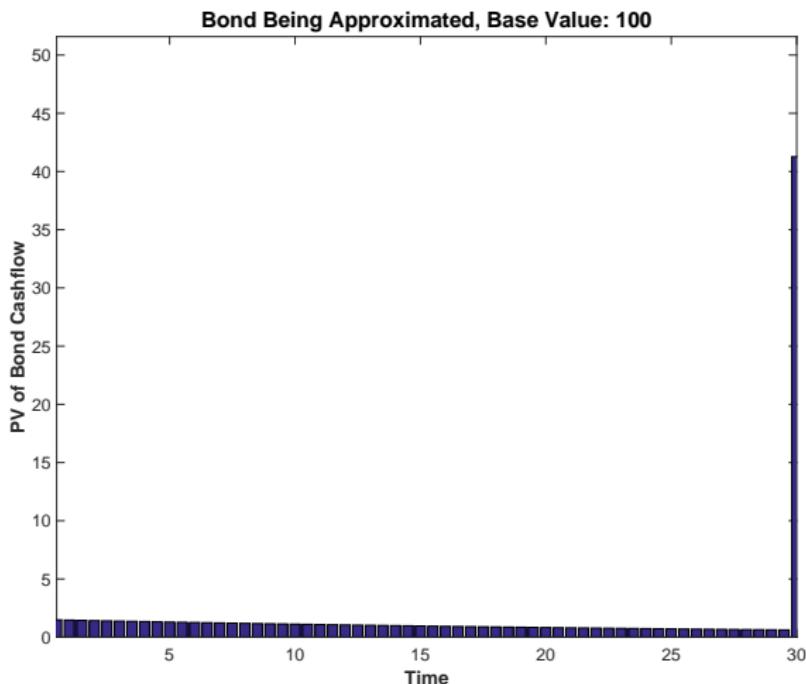


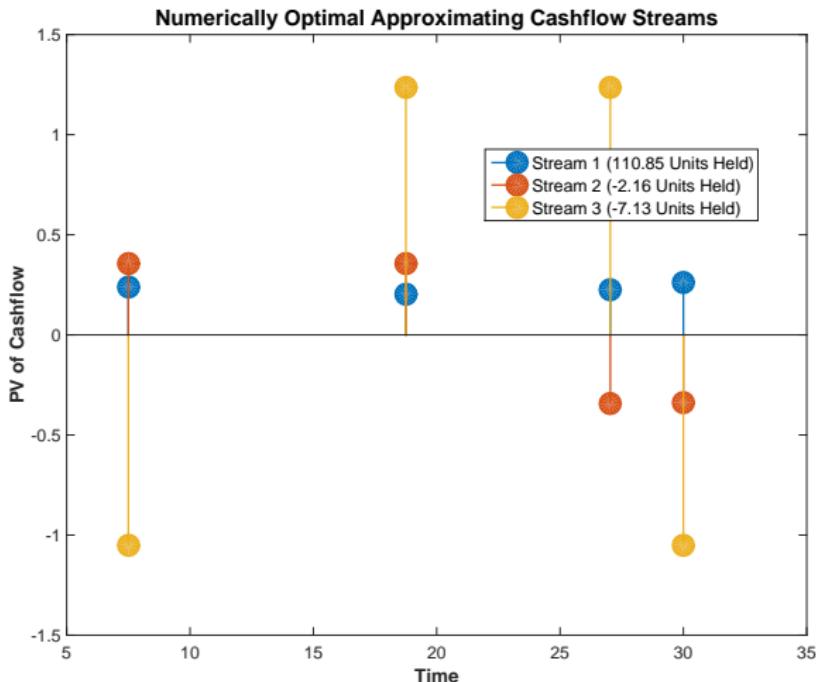


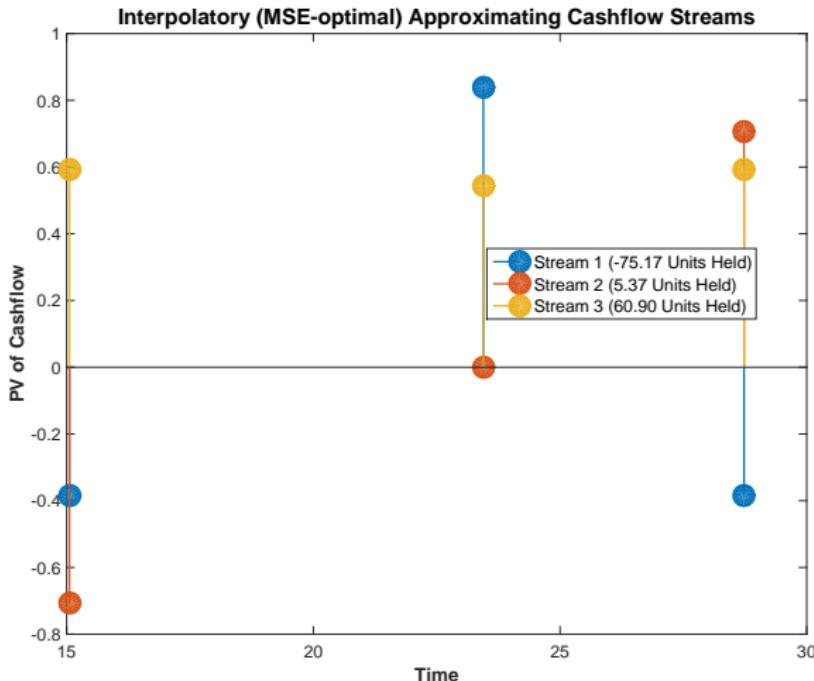


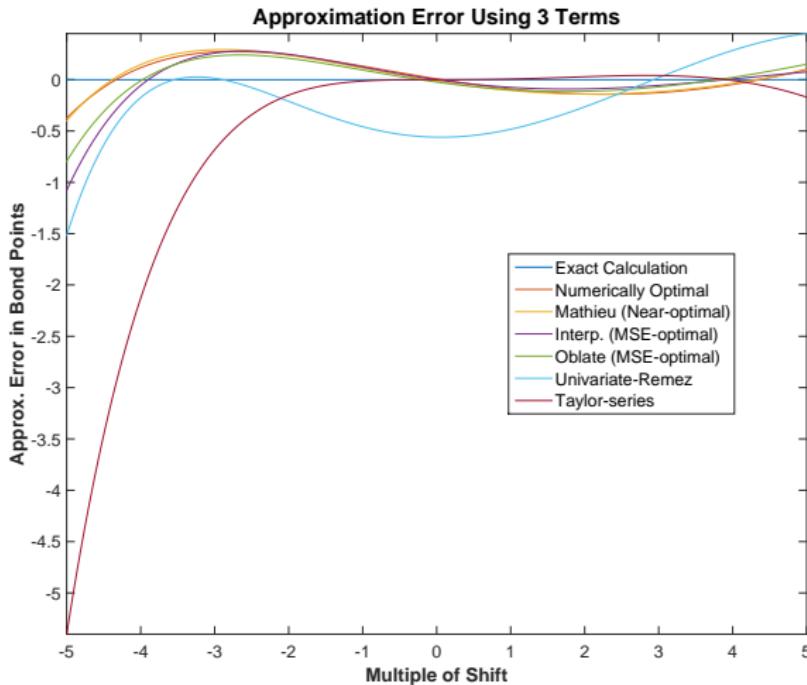


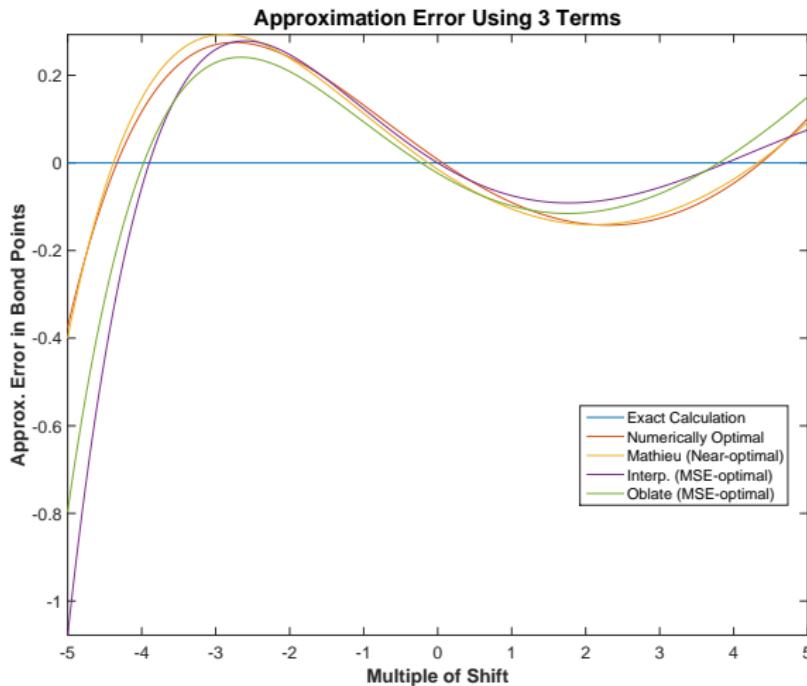


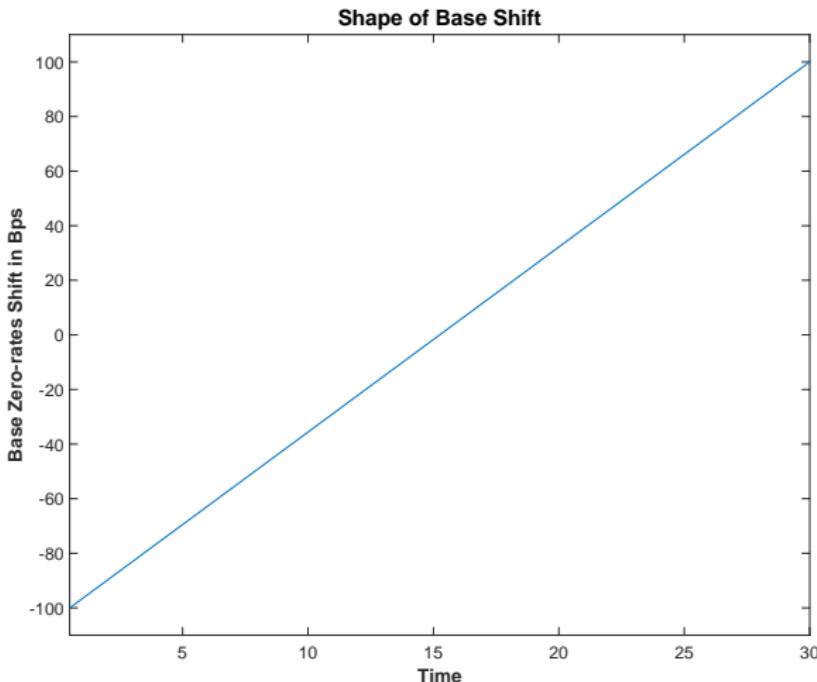


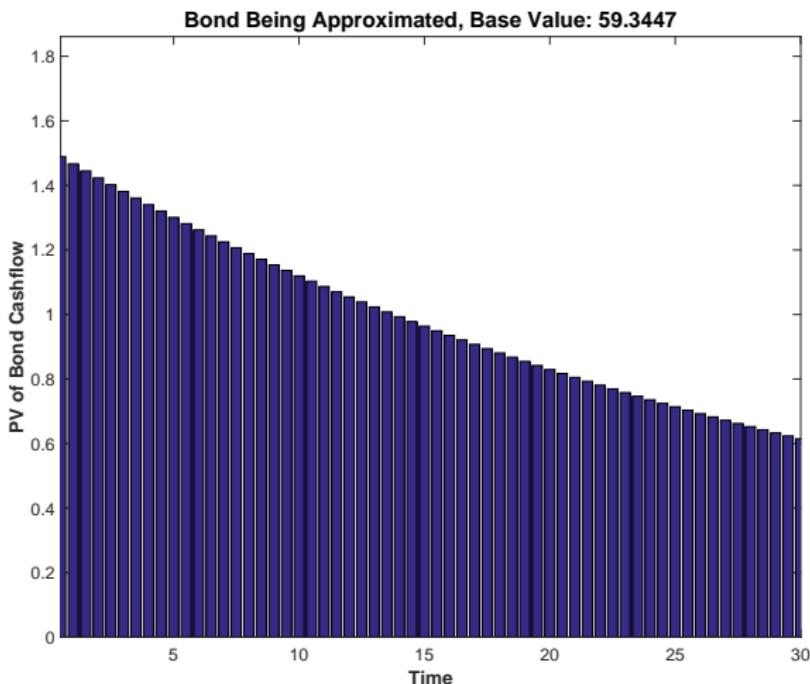


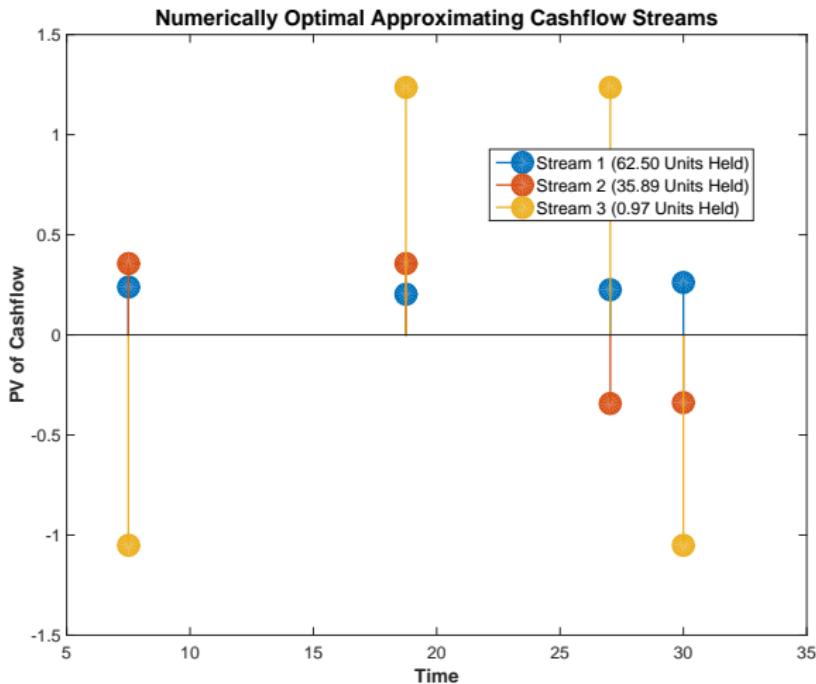


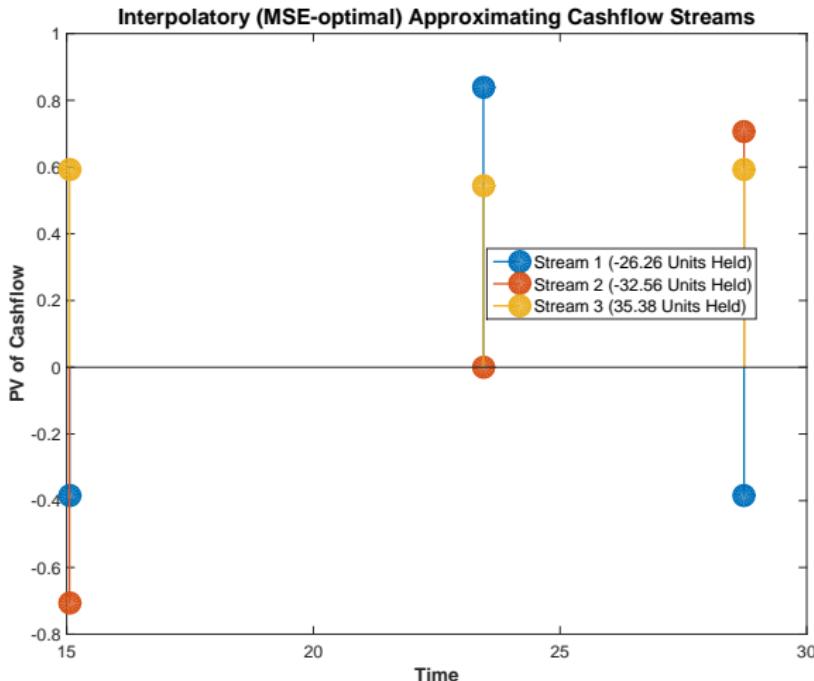


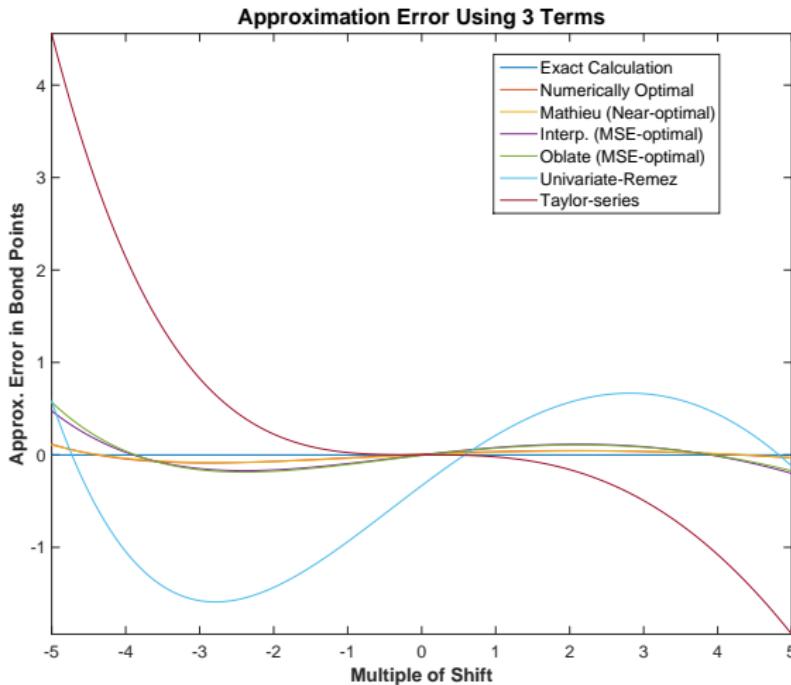


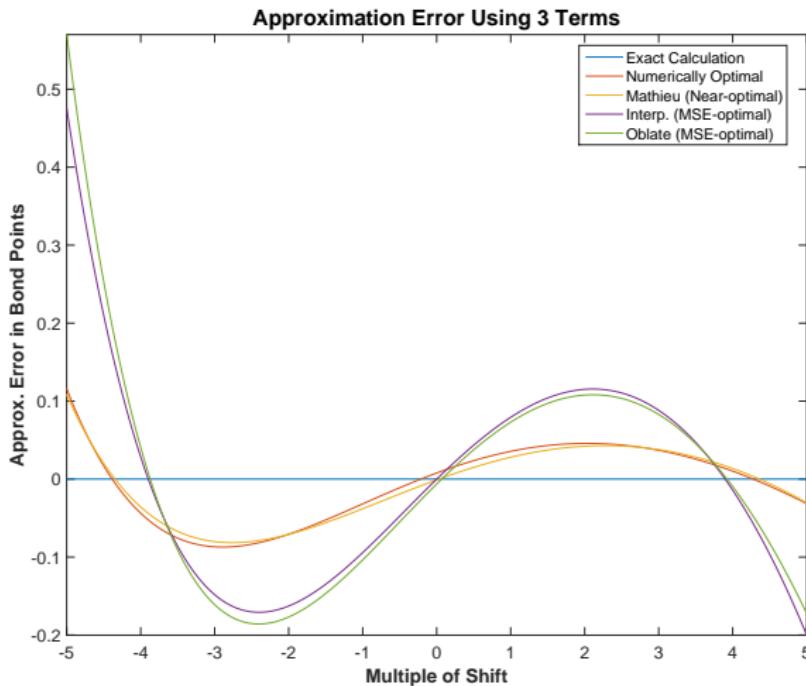


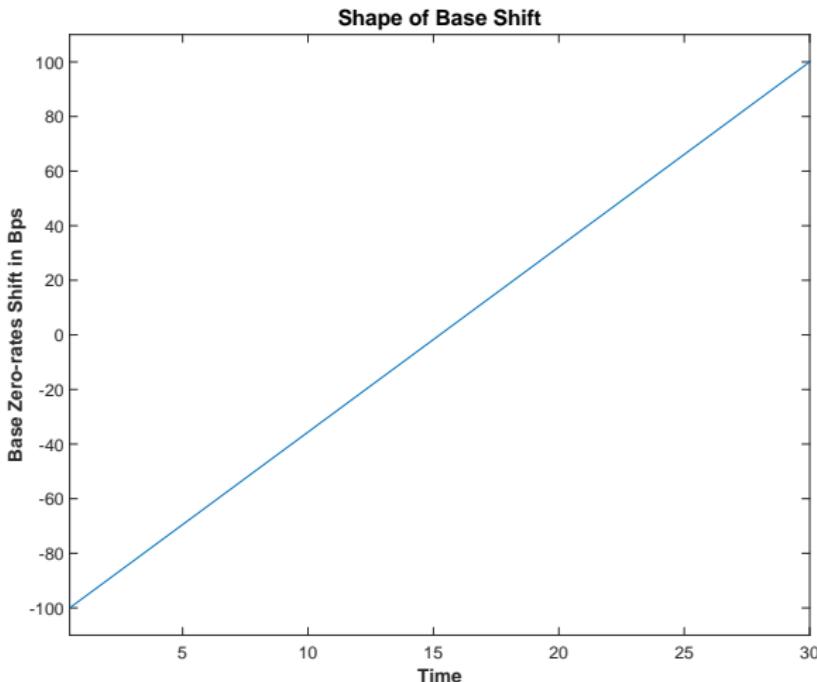


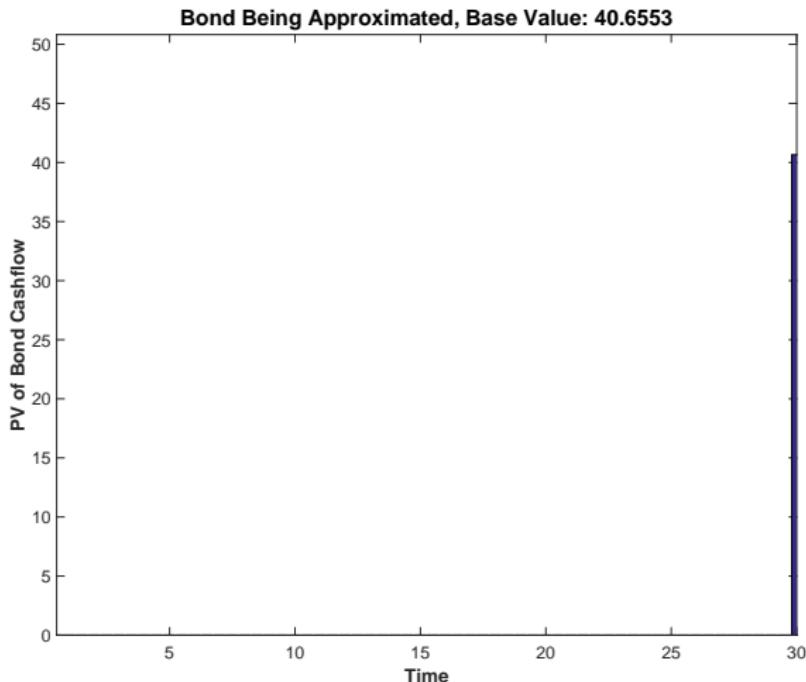


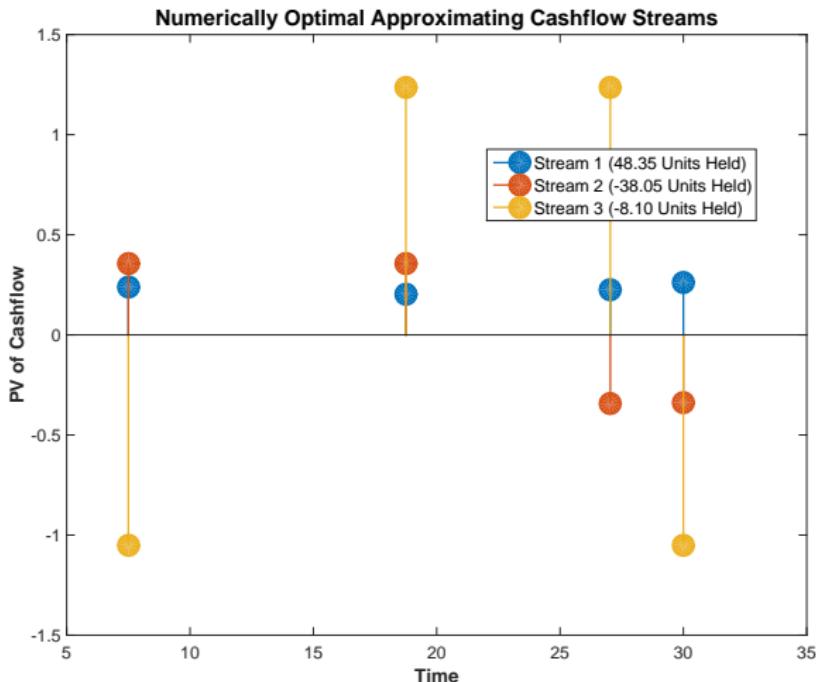


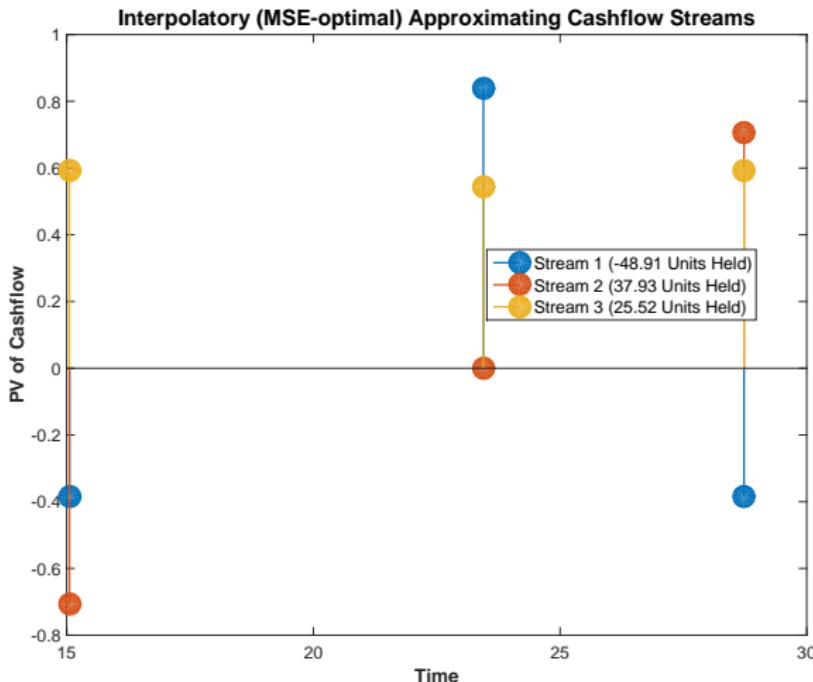


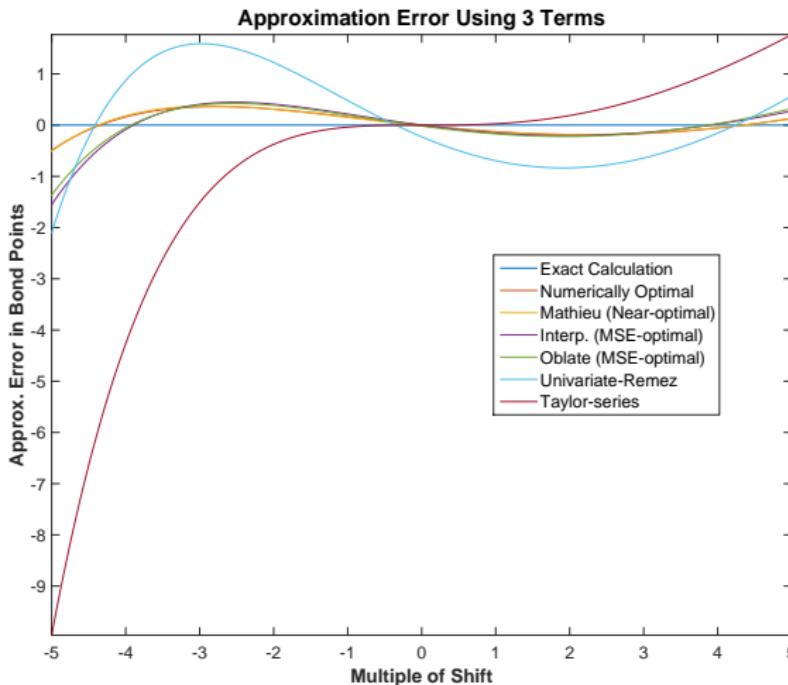


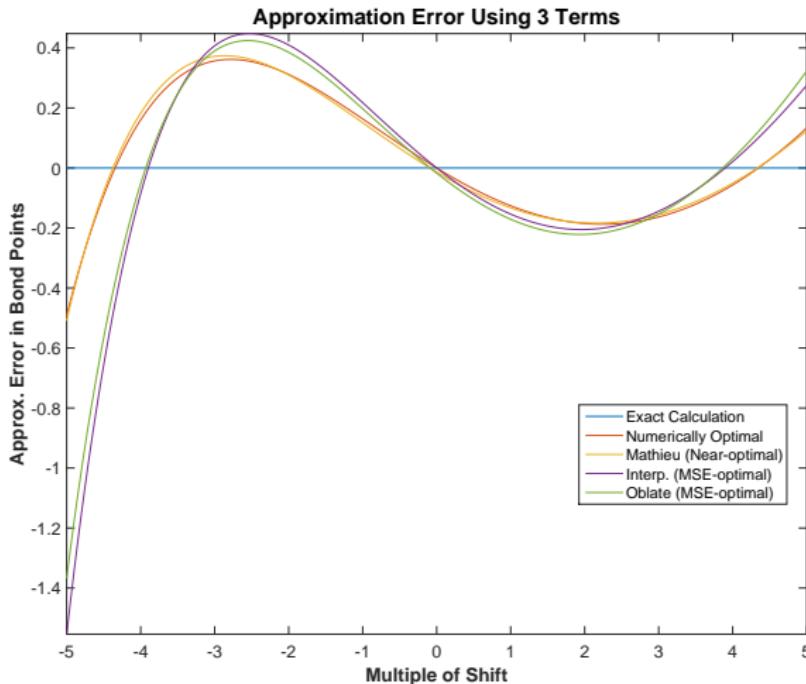


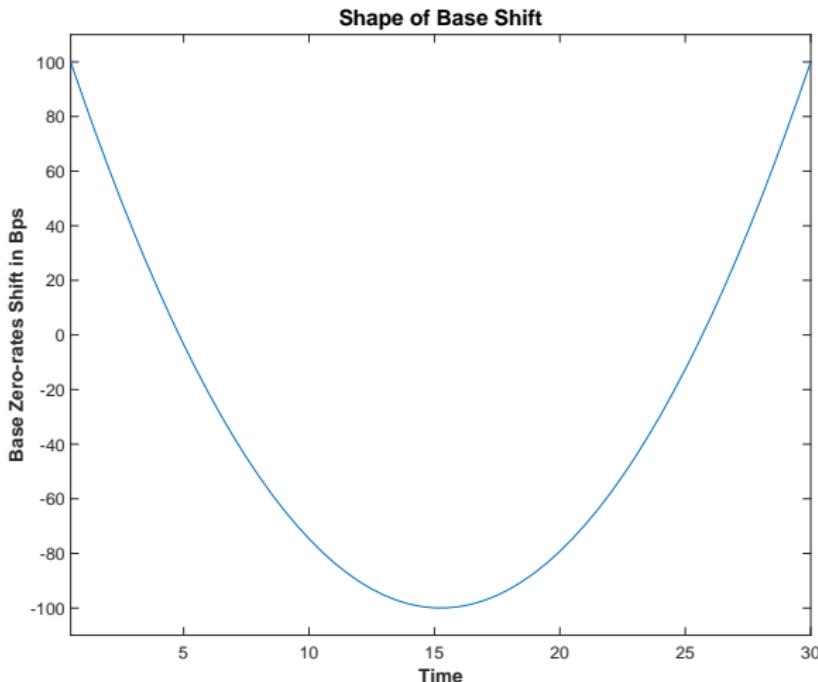


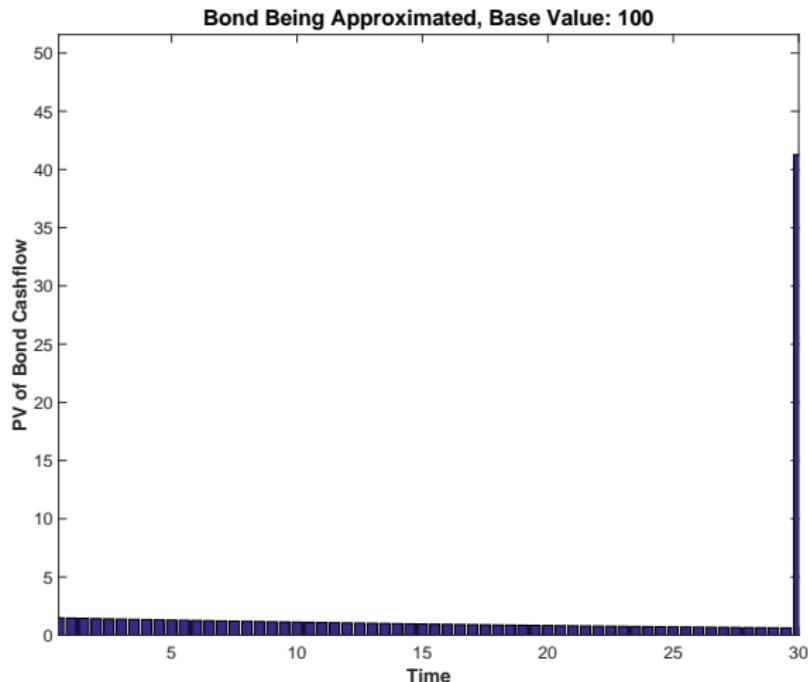


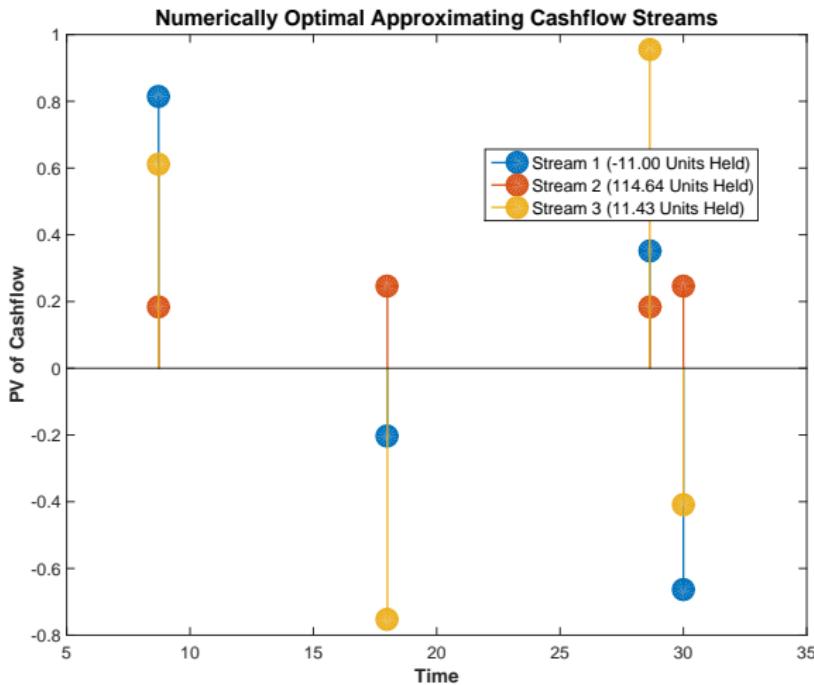


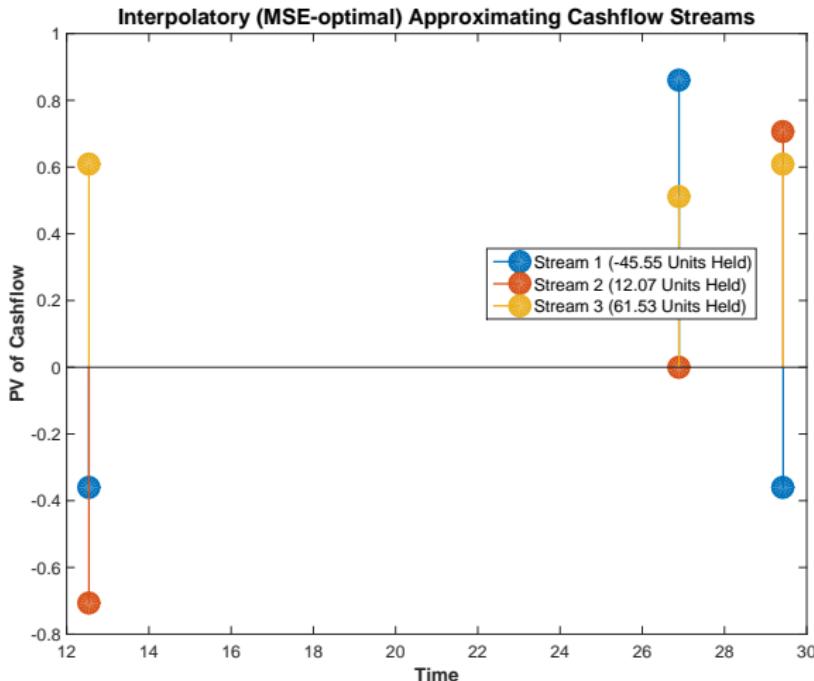


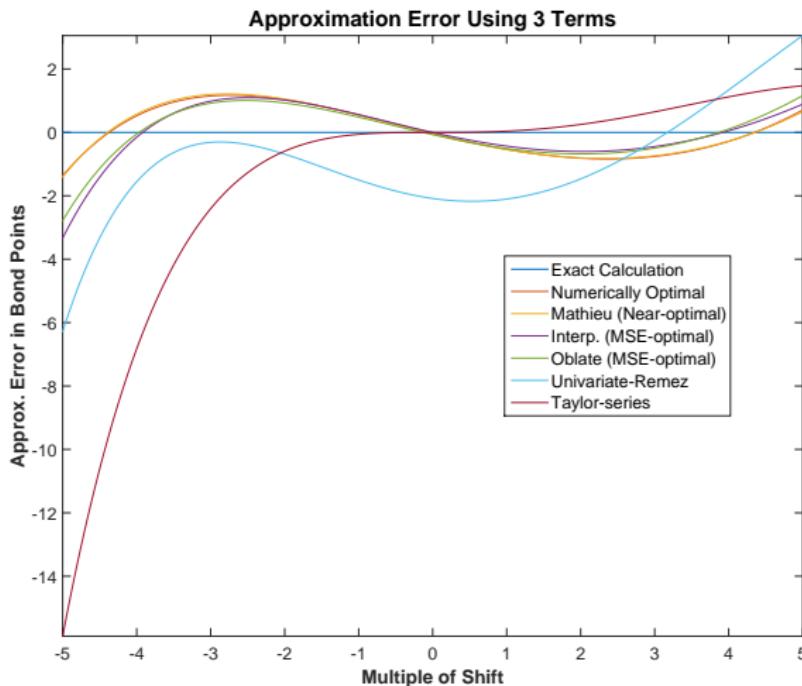


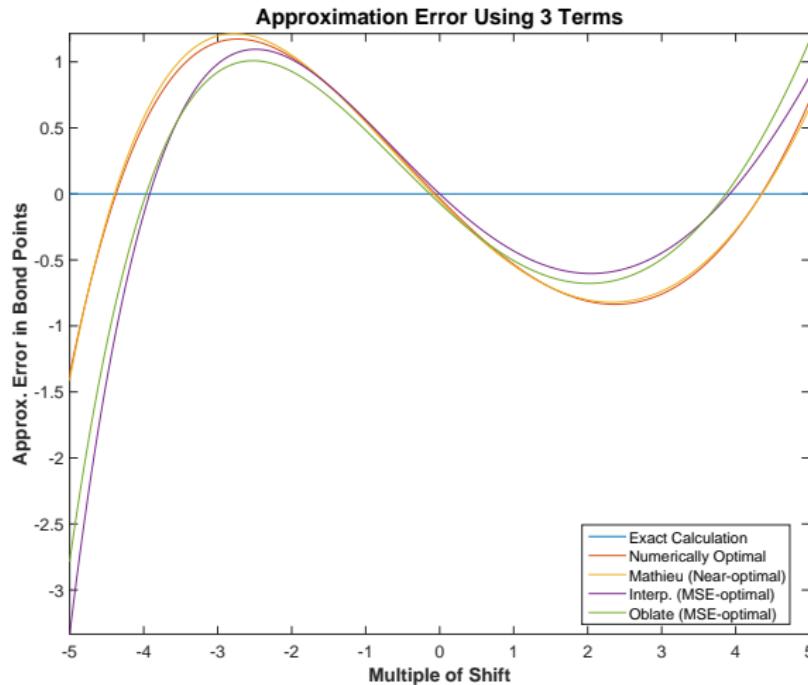


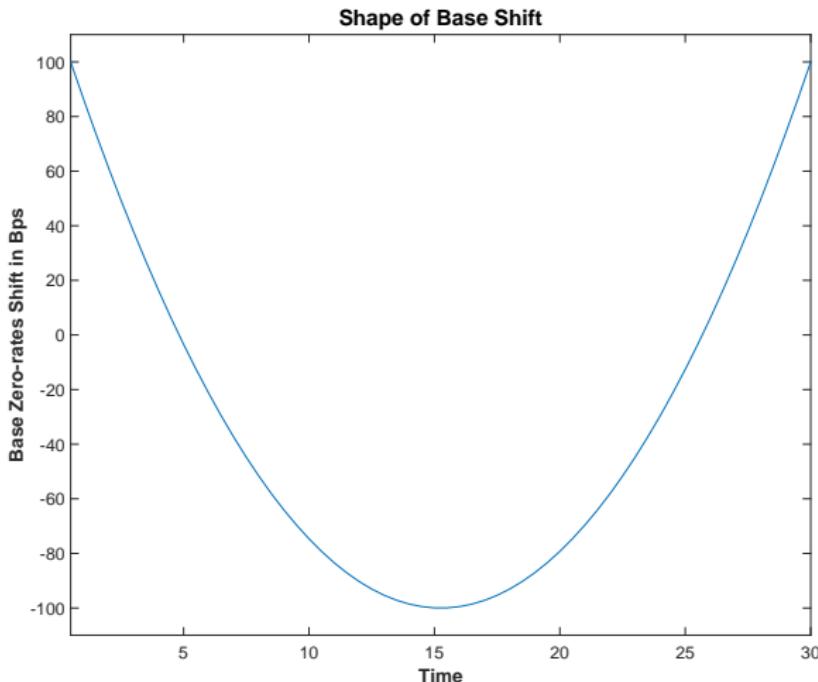


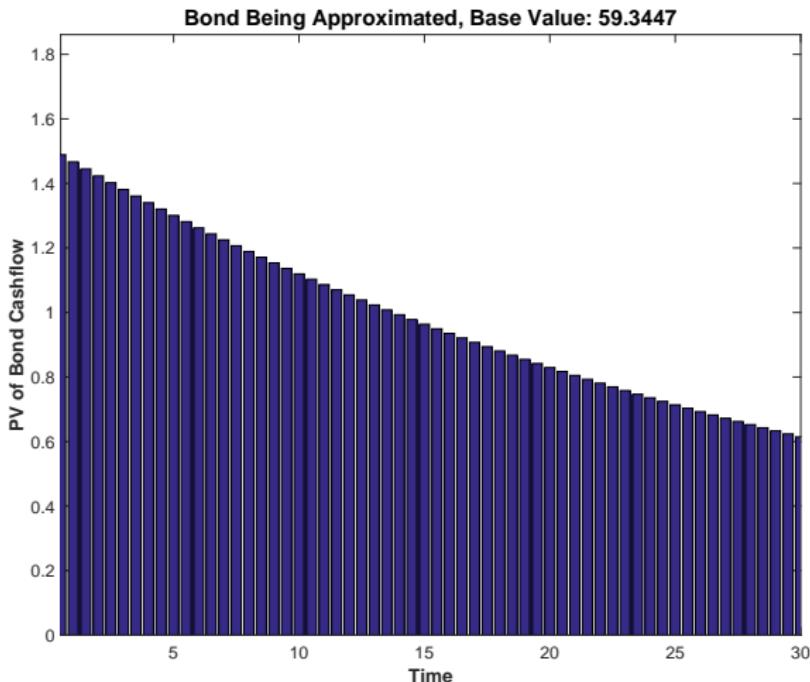


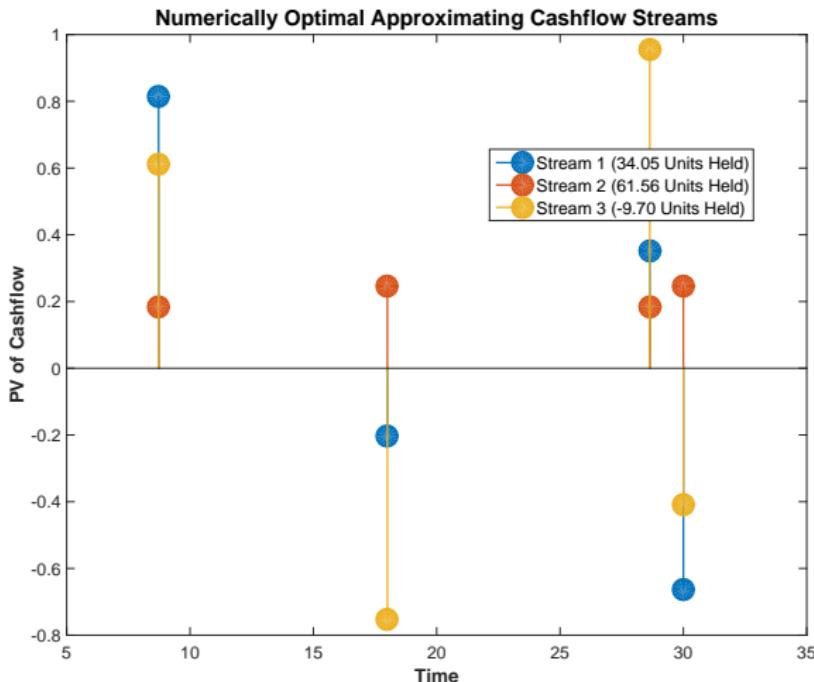


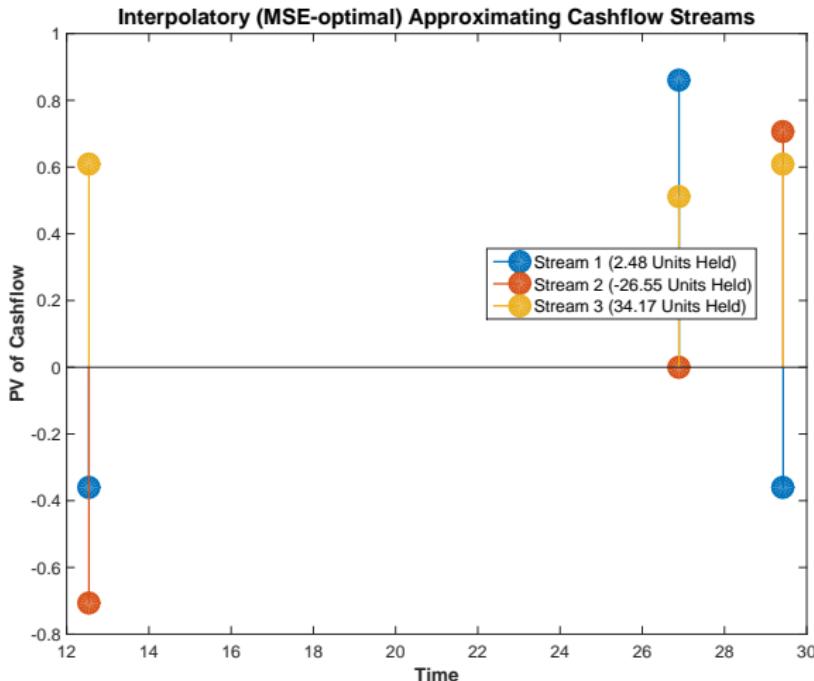


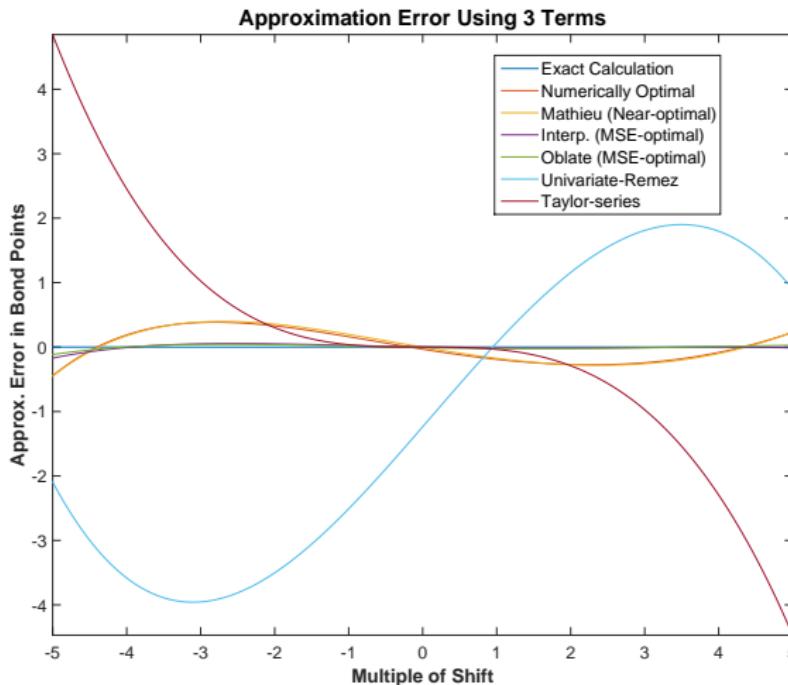


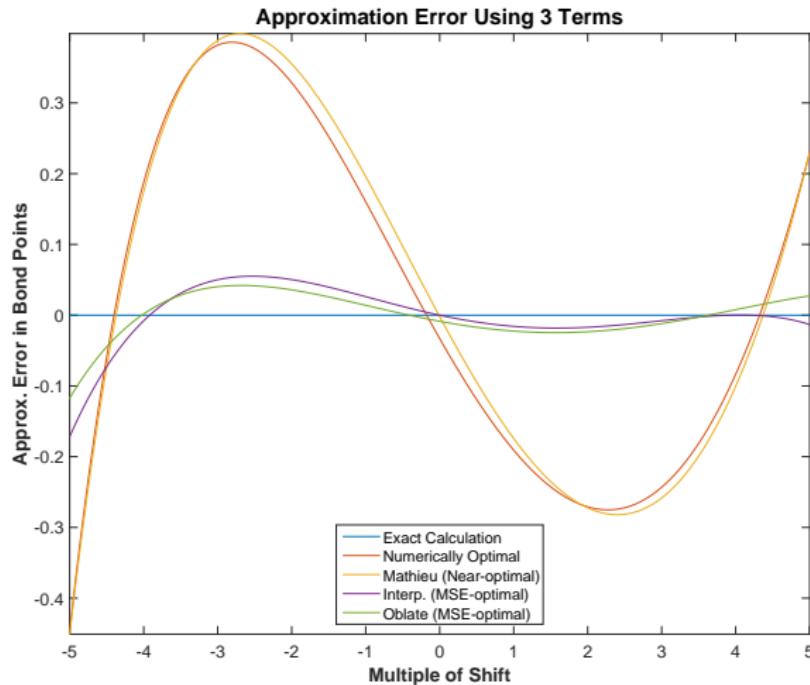


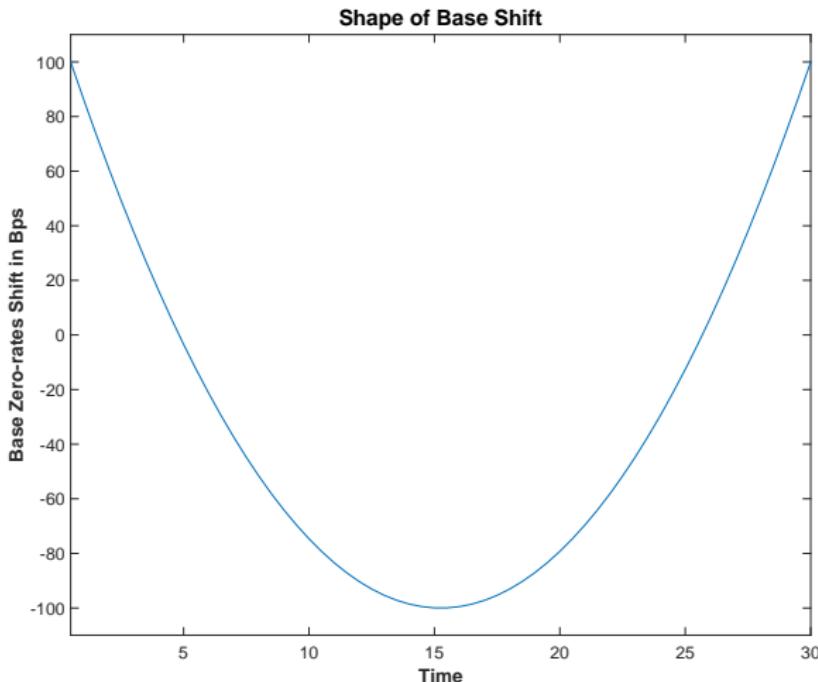


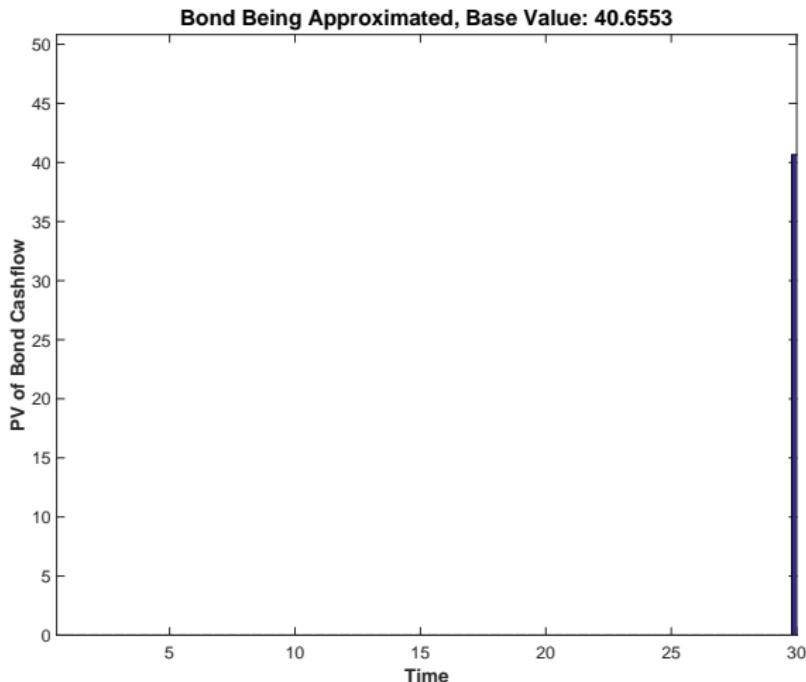


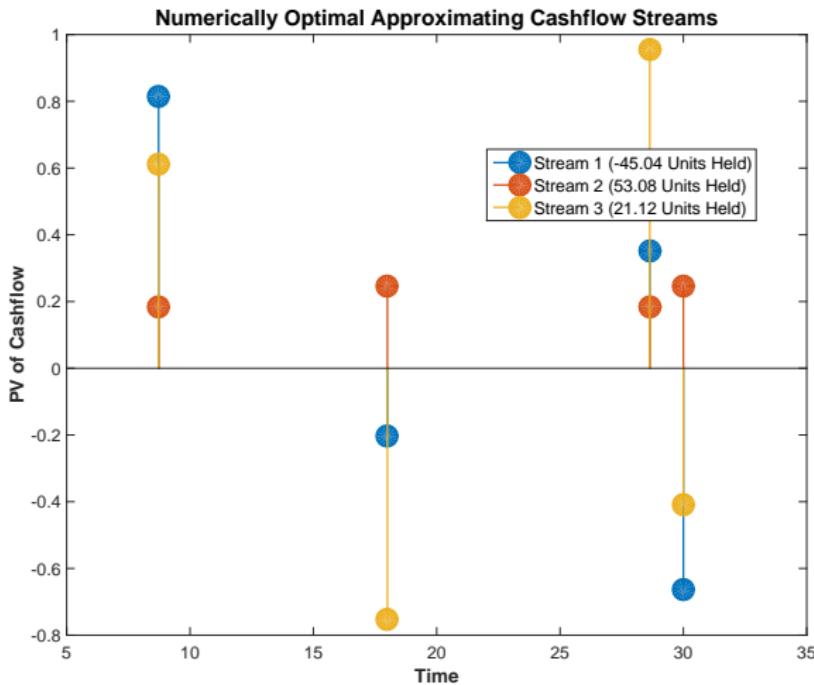


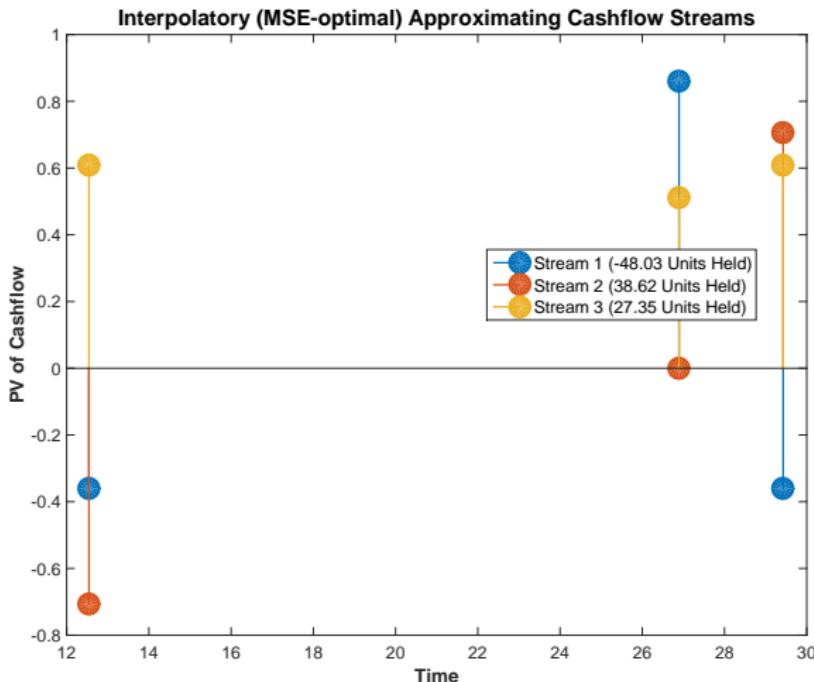


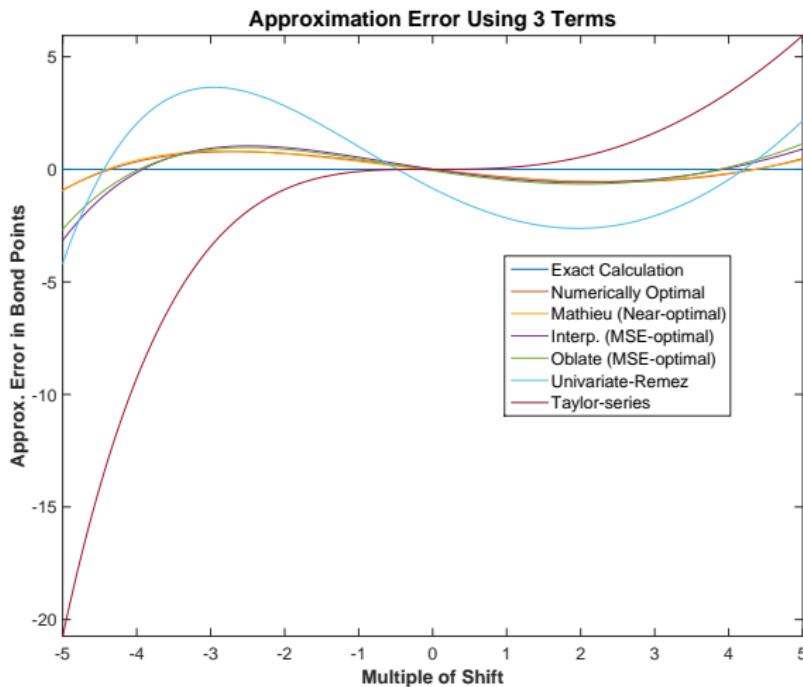


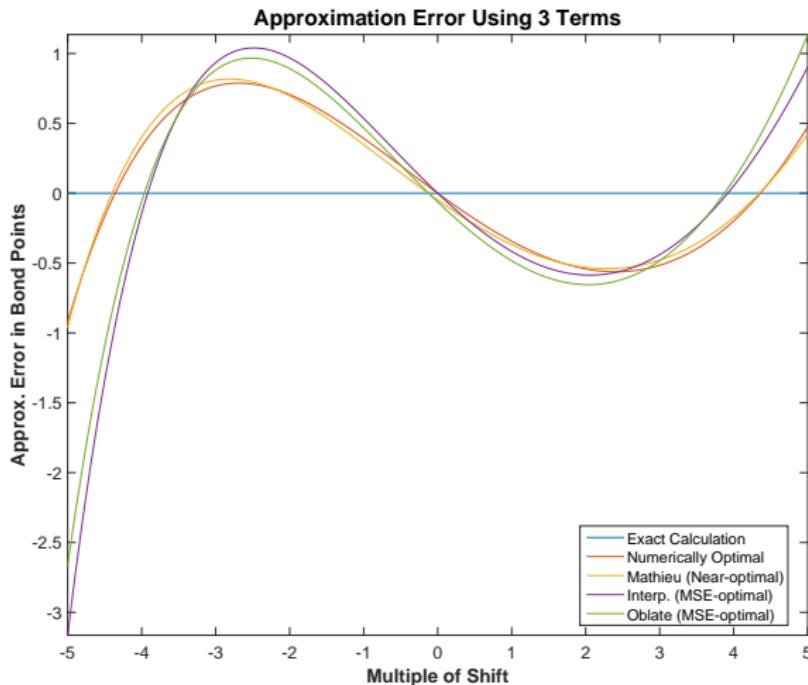


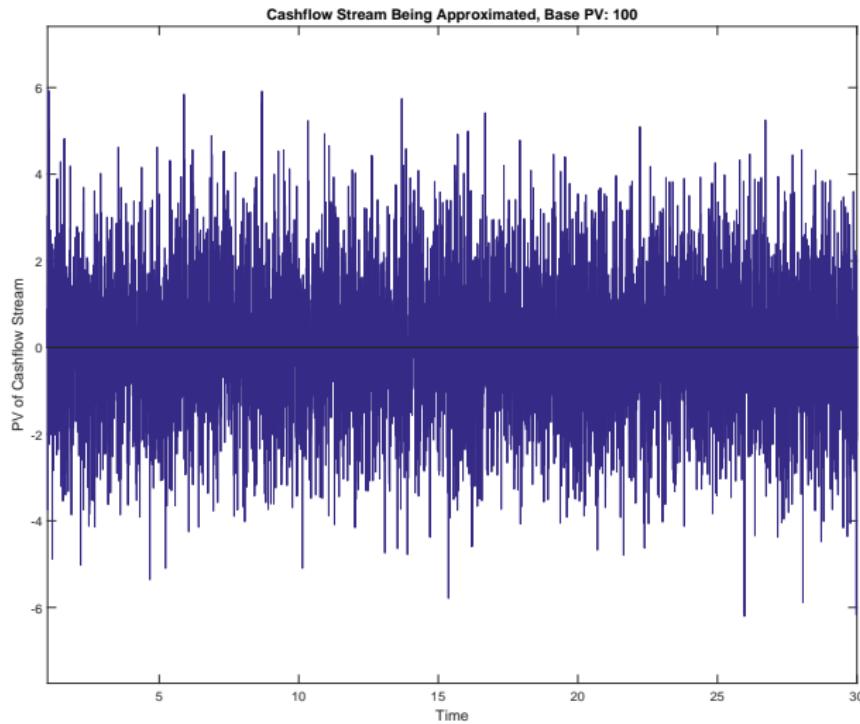












References |

- Anderson, Evan W., Hansen, Lars Peter, & Sargent, Thomas J. 2003. A Quartet of Semigroups for Model Specification, Robustness, Prices of Risk, and Model Detection. *Journal of the European Economic Association*, **1**, 68–123.
- Barndorff-Nielsen, Ole E., & Shephard, Neil. 2001. Non-Gaussian Ornstein-Uhlenbeck-based Models and Some of Their Uses in Financial Economics. *Journal of the Royal Statistical Society: Series B*, **63**, 167–241.
- Borovička, Jaroslav, Hansen, Lars Peter, & Scheinkman, José A. 2014. Misspecified Recovery. *working paper*.

References II

- Borsuk, K. 1933. Drei Sätze über die n -dimensionale euklidische Sphäre. *Fundamenta Mathematicae*, **20**, 177–191.
- Campbell, John Y., Sunderam, Adi, & Viceira, Luis M. 2013. Inflation Bets or Deflation Hedges? The Changing Risks of Nominal Bonds. *working paper*.
- Campbell, John Y., Pflueger, Carolin, & Viceira, Luis M. 2014. Monetary Policy Drivers of Bond and Equity Risks. *NBER Working Paper, 20070*.
- Chamberlain, Gary. 2000. Econometric Applications of Maxmin Expected Utility. *Journal of Applied Econometrics*, **15**, 625–644.

References III

- Chamberlain, Gary. 2001. Minimax Estimation and Forecasting in a Stationary Autoregression Model. *American Economic Review*, **91**, 55–59.
- Cox, John C., Jonathan E. Ingersoll, Jr., & Ross, Stephen A. 1979. Duration and the Measurement of Basis Risk. *The Journal of Business*, **52**, 51–61.
- Cox, John C., Jonathan E. Ingersoll, Jr., & Ross, Stephen A. 1985. A Theory of the Term Structure of Interest Rates. *Econometrica*, **53**, 385–407.
- Duffie, Darrell, Pan, Jun, & Singleton, Kenneth. 2000. Transform Analysis and Asset Pricing for Affine Jump-diffusions. *Econometrica*, **68**, 1343–1376.

References IV

- Greenwood, Robin, & Vayanos, Dimitri. 2014. Bond Supply and Excess Bond Returns. *Review of Financial Studies*, **27**, 663–713.
- Hansen, Lars Peter. 2012. Dynamic Valuation Decomposition within Stochastic Economies. *Econometrica*, **80**, 911–967.
- Hansen, Lars Peter, & Sargent, Thomas J. 1995. Discounted Linear Exponential Quadratic Gaussian Control. *IEEE Transactions on Automatic Control*, **40**, 968–971.
- Hansen, Lars Peter, & Sargent, Thomas J. 2008. *Robustness*. Princeton, NJ: Princeton University Press.
- Hansen, Lars Peter, & Sargent, Thomas J. 2012. Three Types of Ambiguity. *Journal of Monetary Economics*, **59**, 422–445.

References V

- Hansen, Lars Peter, & Scheinkman, José A. 2009. Long Term Risk: An Operator Approach. *Econometrica*, **77**, 177–234.
- Hansen, Lars Peter, Sargent, Thomas J., & Thomas D. Tallarini, Jr. 1999. Robust Permanent Income and Pricing. *Review of Economic Studies*, **66**, 873–907.
- Hansen, Lars Peter, Sargent, Thomas J., Turmuhambetova, Gauhar, & Williams, Noah. 2006. Robust Control and Model Misspecification. *Journal of Economic Theory*, **128**, 45–90.
- Heath, David, Jarrow, Robert, & Morton, Andrew. 1992. Bond Pricing and the Term Structure of Interest Rates: A New Methodology for Contingent Claims Valuation. *Econometrica*, **60**, 77–105.

References VI

- Hicks, John R. 1939. *Value and Capital*. Oxford: Clarendon Press.
- Hodge, D. B. 1970. Eigenvalues and Eigenfunctions of the Spheroidal Wave Equation. *Journal of Mathematical Physics*, **11**, 2308–2312.
- Kolmogorov, A. 1936. Über die beste Annäherung von Funktionen einer gegebenen Funktionenklasse. *Annals of Mathematics*, **37**, 107–110.

References VII

- Macaulay, Frederick R. 1938. *Some Theoretical Problems Suggested by the Movements of Interest Rates, Bond Yields and Stock Prices in the United States Since 1856*. New York: Columbia University Press for the National Bureau of Economic Research. Reprinted by Risk Books as part of the Risk Classics Library, 1999.
- Mairhuber, John C. 1956. On Haar's Theorem Concerning Chebychev Approximation Problems Having Unique Solutions. *Proceedings of the American Mathematical Society*, 7, 609–615.

References VIII

- Melkman, Avraham A., & Micchelli, Charles A. 1978. Spline Spaces Are Optimal for L^2 n -Width. *Illinois Journal of Mathematics*, **22**, 541–564.
- Micchelli, Charles A., & Pinkus, Allan. 1979. The n -Widths of Rank $n + 1$ Kernels. *Journal of Integral Equations*, **1**, 111–130.
- Piazzesi, Monika. 2010. Affine Term Structure Models. In: Aït-Sahalia, Yacine, & Hansen, Lars Peter (eds), *Handbook of Finance*. Oxford: North-Holland.

References IX

- Pinkus, Allan. 1985. *n-Widths in Approximation Theory*.
Ergebnisse der Mathematik und ihrer Grenzgebiete 3. Folge · Band 7. A Series of Modern Surveys in Mathematics. Berlin: Springer-Verlag.
- Redington, F. M. 1952. Review of the Principles of Life-office Valuations. *Journal of the Institute of Actuaries*, **78**, 286–340.
- Ross, Steve. 2014. The Recovery Theorem. *Journal of Finance*, **forthcoming**.
- Samuelson, Paul A. 1945. The Effect of Interest Rate Increases on the Banking System. *The American Economic Review*, **35**, 16–27.

References X

- Tikhomirov, V. M. 1960. Diameters of Sets in Function Spaces and the Theory of Best Approximations. *Russian Mathematical Surveys*, **15**, 75–111.
- Trefethen, Lloyd N. 2013. *Approximation Theory and Approximation Practice*. Other Titles in Applied Mathematics, vol. 128. Philadelphia: Society for Industrial and Applied Mathematics.
- Vasicek, Oldrich A. 1977. An Equilibrium Characterization of the Term Structure. *The Journal of Financial Economics*, **5**, 177–188.

References XI

Zayed, Ahmed I. 2007. A Generalization of the Prolate Spheroidal Wave Functions. *Proceedings of the American Mathematical Society*, **135**, 2193–2203.

Zhang, Shanjie, & Jin, Jianming. 1996. *Computation of Special Functions*. New York: John Wiley & Sons, Inc.