Herb Kunze · Davide La Torre Franklin Mendivil · Edward R. Vrscay

Fractal-Based Methods in Analysis



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In memory of Bruno Forte mentor, teacher, collaborator, friend.

Preface

The idea of modeling the behaviour of phenomena at multiple scales has become a useful tool in both pure and applied mathematics. Fractal-based techniques lie at the heart of this area, as fractals are inherently multiscale objects. Fractals have increasingly become a useful tool in real applications; they very often describe such phenomena better than traditional mathematical models.

Fractal-Based Methods in Analysis draws together, for the first time in book form, methods and results from almost 20 years of research on this topic, including new viewpoints and results in many of the chapters. For each topic, the theoretical framework is carefully explained. Numerous examples and applications are presented.

The central themes are *self-similarity* across scales (exact or approximate) and *contractivity*. In applications, this involves introducing an appropriate space for contractive operators and approximating the "target" mathematical object by the fixed point of one of these contractions. Under fairly general conditions, this approximation can be extremely good. This idea emerged from *fractal image compression*, where an image is encoded by the parameters of a contractive transformation (see Sect. 3.1 and Figs. 3.3 and 3.4). The first step in extending this methodology is to construct interesting contractive operators on many different types of spaces. After this theoretical framework has been established, the next step is to apply the methodology in practical problems. In this book, we present extensive examples of both of these steps.

We originally conceived a document that we could give to our students to help them learn the background for their research. This document would contain an introduction to fractals via iterated function systems (IFSs) and some of the important subsequent developments, all from this IFS viewpoint. This document has since taken on a life of its own that has resulted in this book. The original goal is reflected in the second chapter, which is designed to serve as the basis for a course.

In the first chapter, we give a "bird's-eye" overview of the area, painting with broad brushstrokes to give the reader the feel and philosophy of the approach. We touch on many of the topics and applications, hoping to share our amazement at the breadth of interesting mathematics and applications and to entice the reader into learning more.

In Chapter 2, we present a brief course on the classical topics in the iterated function systems viewpoint on fractals. In order to help the reader who might be seeing the material for the first time, we have included many exercises in this chapter. This is in keeping with our own desire to use this particular chapter as the basis for a course on IFS fractals.

In Chapters 3–5 we carefully develop the IFS framework in a large variety of settings. In particular, in Chapter 3 we develop a theory of IFSs on various spaces of functions, including the interaction of IFSs with integral transforms and IFSs on wavelet spaces. IFSs on spaces of transforms have been used in mathematical imaging, with IFSs on Fourier transforms having applications in magnetic resonance imaging (MRI). Chapter 4 extends this to IFSs on multifunctions (set-valued functions) and measure-valued functions. Again, the primary motivation and application for this framework is in mathematical imaging. Chapter 5 proceeds to a careful construction of the framework in various spaces of measures, with many new results. We consider signed measures, vector measures, and multimeasures (set-valued measures). Furthermore, there is a discussion of "generalized" measures as dual objects to Lipschitz spaces. This is a very useful class of "measures" for many purposes.

In Chapter 6, we turn to another classical topic in IFS fractals, that of ergodic theory and the "chaos game." An IFS defines a dynamical system, which in turn generates an invariant measure. Chapter 6 extends the "chaos game" to IFSs on various function spaces, including random algorithms for generating wavelet analysis and wavelet synthesis.

Chapters 7 and 8 present an extensive range of applications of fractal-based methods to inverse problems. The models that we discuss span the range from ordinary differential equations (ODEs), partial differential equations (PDEs), and random differential equations to stochastic differential equations (SDEs) and more. The range of application topics presented is equally broad, from models in physics to Preface

biological, population, and economic models. These two chapters are just the start of the possible application areas and serve to illustrate the power of fractal-based methods.

The preface of a book is also the customary place for acknowledgments and expressions of thanks to appropriate people – collaborators, students, and individuals who have, in whatever way, helped the authors with their work or understanding of the subject material. In our case, the list of such people is quite long and the risk of omission quite large, so we shall keep our acknowledgments rather brief. First of all (and in a somewhat chronological order). Ed Vrscav would like to thank Michael F. Barnsley who, while at Georgia Tech, introduced him to the fields of fractal geometry and fractal image compression. He would also like to thank Jacques Lévy-Véhel, Dietmar Saupe and Claude Tricot for invaluable discussions, collaborations and assistance that began in the late-1980s and led to the formation of the "Waterloo Fractal Coding and Analysis Project." It was the attraction of Bruno Forte, former Chair of Applied Mathematics, University of Waterloo, to the "Waterloo Project" that contributed to its significant initial growth, in particular the mathematical formulation of generalized fractal transforms and associated inverse problems. Further growth of the project was made possible with the arrival of Franklin Mendivil. After retiring from Waterloo in 1995, Bruno would return to Italy to assume Emeritus Professorships, first at the University of Bari and then at the University of Verona. Here, he would eventually supervise Davide La Torre's Master's thesis on inverse problem for fractal transforms. Davide La Torre would like to thank Vincenzo Capasso and Bruno Forte for addressing him in these topics and for the suggestions, the inspiration, and the support that they have given him during his academic career. Bruno Forte was the first professor Herb Kunze met in the classroom while and undergraduate at the University of Waterloo. A few years later. Herb worked repeatedly as Bruno's teaching assistant for the same advanced Calculus course. Herb Kunze wishes to acknowledge the influential impact that Bruno had on his early academic life.

For these reasons we are dedicating this book to him.

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Guelph, Ontario Milan, Italy Wolfville, Nova Scotia Waterloo, Ontario Herb Kunze Davide La Torre Franklin Mendivil Edward R.Vrscay



Contents

 1.1 Fractal transforms and self-similarity 1.2 Self-similarity: A brief historical review	is"?	1
 1.2 Self-similarity: A brief historical review		3
 1.2.1 The construction of self-similar sets 1.2.2 The construction of self-similar measures 1.3 Induced fractal transforms 1.4 Inverse problems for fractal transforms and "collage coding" 2 Basic IFS Theory 2.1 Contraction mappings and fixed points 2.1.1 Inverse problem 2.2 Iterated Function System (IFS) 2.2.1 Motivating example: The Cantor set 2.2.2 Space of compact subsets and the Hausdorff metric 2.2.3 Definition of IFS 2.2.4 Collage theorem for IFS 2.2.5 Continuous dependence of the attractor . 2.4 The chaos game 2.5 IFS with probabilities 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M⁴ 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions		8
 1.2.2 The construction of self-similar measures 1.3 Induced fractal transforms		8
 1.3 Induced fractal transforms		11
 1.4 Inverse problems for fractal transforms and "collage coding"		12
 "collage coding"		
 1.4.1 Fractal image coding		16
 2 Basic IFS Theory 2.1 Contraction mappings and fixed points 2.1.1 Inverse problem 2.2 Iterated Function System (IFS) 2.2.1 Motivating example: The Cantor set 2.2.2 Space of compact subsets and the Hausdorff metric 2.2.3 Definition of IFS 2.2.4 Collage theorem for IFS 2.2.5 Continuous dependence of the attractor 2.3 Code space and the address map 2.4 The chaos game 2.5 IFS with probabilities 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M⁴ 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions 2.6.1 IFS with condensation 		18
 2.1 Contraction mappings and fixed points 2.1.1 Inverse problem		21
 2.1.1 Inverse problem		21
 2.2 Iterated Function System (IFS) 2.2.1 Motivating example: The Cantor set 2.2.2 Space of compact subsets and the Hausdorff metric 2.2.3 Definition of IFS 2.2.4 Collage theorem for IFS 2.2.5 Continuous dependence of the attractor . 2.3 Code space and the address map 2.4 The chaos game 2.5 IFS with probabilities 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M⁴ 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions 2.6 For etablic terms of the invariant 		25
 2.2.1 Motivating example: The Cantor set 2.2.2 Space of compact subsets and the Hausdorff metric		27
 2.2.2 Space of compact subsets and the Hausdorff metric		27
Hausdorff metric 2.2.3 Definition of IFS 2.2.4 Collage theorem for IFS 2.2.5 Continuous dependence of the attractor 2.3 Code space and the address map 2.4 The chaos game 2.5 IFS with probabilities 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M [*] 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions 2.6.1 IFS with condensation		
 2.2.3 Definition of IFS 2.2.4 Collage theorem for IFS 2.2.5 Continuous dependence of the attractor . 2.3 Code space and the address map 2.4 The chaos game 2.5 IFS with probabilities 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M⁴ 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions 2.6.1 IFS with condensation 		30
 2.2.4 Collage theorem for IFS 2.2.5 Continuous dependence of the attractor . 2.3 Code space and the address map 2.4 The chaos game 2.5 IFS with probabilities 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M⁴ 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions 2.6.1 IFS with condensation 		34
 2.2.5 Continuous dependence of the attractor . 2.3 Code space and the address map 2.4 The chaos game 2.5 IFS with probabilities 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M⁴ 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions		39
 2.3 Code space and the address map 2.4 The chaos game 2.5 IFS with probabilities 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M[*] 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions		40
 2.4 The chaos game		43
 2.5 IFS with probabilities		48
 2.5.1 IFSP and invariant measures 2.5.2 Moments of the invariant measure and M[*] 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions 2.6.1 IFS with condensation 		51
 2.5.2 Moments of the invariant measure and M[*] 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions 2.6.1 IFS with condensation 		51
 2.5.3 The ergodic theorem for IFSP 2.6 Some classical extensions 2.6.1 IFS with condensation 2.6.2 Englished by the englished for the englished of the englis	\mathbb{I}^*	62
2.6 Some classical extensions 2.6.1 IFS with condensation		65
2.6.1 IFS with condensation		69
		70
2.0.2 Fractal interpolation functions		72

		2.6.3	Graph-directed IFS	74 79
		2.0.4	It is with mininely many maps	15
3	IFS	on S	paces of Functions	87
	3.1	Motiv	vation: Fractal imaging	87
	3.2	IFS o	n functions	92
		3.2.1	Uniformly contractive IFSM	92
		3.2.2	IFSM on $\mathcal{L}^p(\mathbb{X})$	95
		3.2.3	Affine IFSM	98
		3.2.4	IFSM with infinitely many maps	99
		3.2.5	Progression from geometric IFS to IFS on	
			functions	100
	3.3	IFS o	n wavelets	102
		3.3.1	Brief wavelet introduction	103
		3.3.2	IFS operators on wavelets (IFSW)	105
		3.3.3	Correspondence between IFSW and IFSM	107
	3.4	IFS a	nd integral transforms	111
		3.4.1	Fractal transforms of integral transforms	113
		3.4.2	Induced fractal operators on fractal transforms	114
		3.4.3	The functional equation for the kernel	116
		3.4.4	Examples	119
4	IFS	, Mul	tifunctions, and Measure-Valued Functions.	125
	4.1	IMS a	and IMS with probabilities	125
		4 1 1		
		4.1.1	Code space	129
	4.2	4.1.1 Iterat	Code space ed function systems on multifunctions	$\begin{array}{c} 129 \\ 130 \end{array}$
	4.2	4.1.1 Iterat 4.2.1	Code space ed function systems on multifunctions Spaces of multifunctions	129 130 130
	4.2	4.1.1 Iterat 4.2.1 4.2.2	Code spaceed function systems on multifunctionsSpaces of multifunctionsSome IFS operators on multifunctions (IFSMF)	129 130 130 132
	4.2	4.1.1 Iterat 4.2.1 4.2.2 4.2.3	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding	129 130 130 132 135
	4.24.3	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images	129 130 130 132 135 140
	4.24.3	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images A fractal transform operator on	129 130 130 132 135 140
	4.24.3	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images A fractal transform operator on measure-valued images	129 130 130 132 135 140 141
	4.24.3	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1 4.3.2	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images A fractal transform operator on measure-valued images Moment relations induced by the fractal	129 130 130 132 135 140 141
	4.24.3	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1 4.3.2	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images A fractal transform operator on measure-valued images Moment relations induced by the fractal transform operator	129 130 130 132 135 140 141 145
5	4.2 4.3 IFS	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1 4.3.2	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images A fractal transform operator on measure-valued images Moment relations induced by the fractal transform operator paces of Measures	129 130 130 132 135 140 141 145 149
5	4.2 4.3 IFS 5.1	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1 4.3.2 on S Signe	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images A fractal transform operator on measure-valued images	$129 \\ 130 \\ 130 \\ 132 \\ 135 \\ 140 \\ 141 \\ 145 \\ 149 \\ 150 \\ 150 \\ 129 \\ 150 \\ 129 \\ 130 \\ 140 \\ 141 \\ 145 \\ 150 $
5	4.24.3IFS 5.1	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1 4.3.2 5 on S Signe 5.1.1	Code space ed function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images A fractal transform operator on measure-valued images	$129 \\ 130 \\ 130 \\ 132 \\ 135 \\ 140 \\ 141 \\ 145 \\ 149 \\ 150 \\ 151 \\ 151 \\ 129 \\ 150 \\ 151 \\ 129 \\ 120 $
5	4.24.3IFS 5.1	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1 4.3.2 5 on S Signe 5.1.1 5.1.2	Code space	$129 \\ 130 \\ 130 \\ 132 \\ 135 \\ 140 \\ 141 \\ 145 \\ 149 \\ 150 \\ 151 \\ 153 \\ 153 \\ 153 \\ 129 \\ 150 \\ 151 \\ 153 \\ 150 \\ 151 \\ 153 \\ 150 \\ 151 \\ 153 \\ 150 \\ 151 \\ 153 \\ 150 $
5	4.2 4.3 IFS 5.1	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1 4.3.2 5 on S Signe 5.1.1 5.1.2 5.1.3	Code space	$129 \\ 130 \\ 130 \\ 132 \\ 135 \\ 140 \\ 141 \\ 145 \\ 149 \\ 150 \\ 151 \\ 153 \\ $
5	4.24.3IFS 5.1	4.1.1 Iterat 4.2.1 4.2.2 4.2.3 Iterat 4.3.1 4.3.2 6 on S Signet 5.1.1 5.1.2 5.1.3	Code space red function systems on multifunctions Spaces of multifunctions Some IFS operators on multifunctions (IFSMF) . An application to fractal image coding ed function systems on measure-valued images A fractal transform operator on measure-valued images	$129 \\ 130 \\ 132 \\ 135 \\ 140 \\ 141 \\ 145 \\ 149 \\ 150 \\ 151 \\ 153 \\ 156 \\ 156$

	5.2	Vecto	r-valued measures	163
		5.2.1	Complete space of vector measures	166
		5.2.2	IFS on vector measures	168
		5.2.3	Coloured fractals	175
		5.2.4	Line integrals on fractal curves	179
		5.2.5	Generalized vector measures	182
		5.2.6	Green's theorem for planar domains with	
			fractal boundaries	183
		5.2.7	Some generalizations for vector measures	186
	5.3	Set-va	alued measures	190
		5.3.1	Complete space of multimeasures	193
		5.3.2	IFS operators on multimeasures	196
		5.3.3	Generalizations for spaces of multimeasures	200
		5.3.4	Union-additive multimeasures	202
		5.3.5	IFS on union-additive multimeasures	206
		5.3.6	Generalities on union-additive multimeasures	209
		5.3.7	Extension of finitely union-additive multimeasures	212
6	The	• Cha	os Game	213
Ū	6.1	Chaos	s game for IFSM	214
	0	6.1.1	Chaos game for nonoverlapping IFSM	214
		6.1.2	Chaos game for overlapping IFSM	217
	6.2	Chaos	s game for wavelets	221
		6.2.1	Rendering a compactly supported	
			scaling function	222
		6.2.2	Modified chaos game algorithm for	
			wavelet generation	224
		6.2.3	Chaos game for wavelet analysis	226
		6.2.4	Chaos game for wavelet synthesis	228
		6.2.5	Some extensions	230
	6.3	Chaos	s game for multifunctions and multimeasures	232
		6.3.1	Chaos game for fractal measures with	
			fractal densities	232
		6.3.2	Chaos game for multifunctions	234
		6.3.3	Chaos game for multimeasures	239
7	Inv	erse F	Problems and Fractal-Based Methods	243
	7.1	Ordin	nary differential equations	244
		7.1.1	Inverse problem for ODEs	248
		7.1.2	Practical Considerations and examples	250
		7.1.3	Multiple, partial, and noisy data sets	259
	7.2	Two-j	point boundary value problems	265
		-		

		7.2.1	Inverse problem for two-point BVPs	268
		7.2.2	Practical considerations and examples	269
	7.3	Quasi	ilinear partial differential equations	276
		7.3.1	Inverse problems for traffic and fluid flow	282
	7.4	Uriso	n integral equations	284
		7.4.1	Inverse problem for Urison integral equations	286
	7.5	Hamr	nerstein integral equations	289
		7.5.1	Inverse problem for Hammerstein	
			integral equations	291
	7.6	Rand	om fixed-point equations	295
		7.6.1	Inverse problem for random DEs	296
	7.7	Stoch	astic differential equations	302
		7.7.1	Inverse problem for SDEs	303
	7.8	Appli	cations	303
		7.8.1	Population dynamics	303
		7.8.2	mRNA and protein concentration	305
		7.8.3	Bacteria and amoeba interaction	307
		7.8.4	Tumor growth	309
8	Fur	ther l	Developments and Extensions	315
0	81	Gener	ralized collage theorems for PDEs	315
	0.1	8.1.1	Elliptic PDEs	318
		8.1.2	Parabolic PDEs	332
		8.1.3	Hyperbolic PDEs	334
		8.1.4	An application: A vibrating string driven by a	
			stochastic process	337
	8.2	Self-s	imilar objects in cone metric spaces	341
		8.2.1	Cone metric space	341
		8.2.2	Scalarizations of cone metrics	343
		8.2.3	Cone with empty interior	349
		8.2.4	Applications to image processing	350
	m.	1		957
A		Soto	cal and Metric Spaces	357
	A.1	Sets .	logical graded	007 050
	A.2	10p0i	Basic definitions	250
		A.2.1	Convergence countability and convertion avient	250
		A.2.2	Convergence, countability, and separation axioms	009 961
		Π.2.0 Δ 9 Λ	Continuity and connectedness	361
	٨э	n.2.4 Motri		365
	л.э		Sequences in metric spaces	363
		A.3.1	Bounded totally bounded and compact sets	264
		A.J.Z	Dounded, totally bounded, and compact sets	304