On the use of stochastic systems for sensing and security

by

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Contents

Content	is i	ii
Abstract	t	ix
Stateme	ent of Originality	xi
Acknow	ledgments xi	iii
Convent	tions	(V
Publicat	ions xv	ʻii
List of F	igures x	ix
List of T	ables	(V
Chapter	1. Introduction	1
1.1	Introduction	2
1.2	Thesis outline	2
1.3	Motivation	3
1.4	Background	4
	1.4.1 System identification	4
	1.4.2 Cryptography	5
1.5	Original contributions	12
Chapter	2. Decisions, Failures, and Stochastic Systems 1	3
2.1	Introduction	4
2.2	Analysis of a biased coin	15
2.3	A hypothetical Roman pot	.6
	2.3.1 Formal model	17

		2.3.2	Analysis of the pot origin distribution	17
	2.4	The rel	iability of identity parades	19
	2.5	Ancien	t judicial procedure	22
	2.6	The rel	iability of cryptographic systems	24
		2.6.1	Code changes caused by memory errors	24
		2.6.2	The effect of memory errors on confidence	26
	2.7	Discus	sion	28
	2.8	Conclu	usion	29
		2.8.1	Original contributions	30
Cł	napter	3. No	nlinear Sensing	31
	3.1	Introdu	uction	32
		3.1.1	Why sense in a nonlinear regime?	32
	3.2	Charac	terisation of nonlinearity in metrology	32
		3.2.1	Direct response measurement	33
		3.2.2	Histogram measurement	33
	3.3	Lineari	sation by noise measurement	34
		3.3.1	Method	34
		3.3.2	Estimation of the derivative	35
		3.3.3	Harmonic distortion	36
		3.3.4	Static error	36
	3.4	Optim	isation for real-time use	39
		3.4.1	Implementation	41
	3.5	Adapta	tion for resource-constrained environments	46
		3.5.1	Results	49
	3.6	Conclu	usion	51
		3.6.1	Original contributions	51
Cł	napter	4. No	ise-based Communication	53
	4.1	Key est	ablishment	54
	4.2	Securit	y definitions	54

4.3	Classi	cal key establishment protocols	56
	4.3.1	Diffie-Hellman	56
	4.3.2	RSA public-key encryption	57
	4.3.3	Shamir's three-pass protocol	58
4.4	A phy	sical implementation of the Shamir three-pass protocol	60
	4.4.1	The mutual information rate between endpoints	61
	4.4.2	Limitations	62
	4.4.3	Experimental Round-Trip Measurements	63
	4.4.4	Demonstration System	64
4.5	The K	ish key distribution system	66
4.6	Attack	king KKD with wave measurement	. 68
	4.6.1	Experimental apparatus	, 70
	4.6.2	Circuit analysis	, 72
	4.6.3	Statistical processing	73
	4.6.4	Experimental results	75
	4.6.5	Proposed countermeasures and alternative explanations	75
	4.6.6	Discussion	. 82
4.7	Attack	king KKD with propagation sensing	. 82
	4.7.1	Quantification of attack effectiveness	. 82
	4.7.2	Nonidealities in the lumped model	. 87
	4.7.3	Transient attacks	. 88
	4.7.4	Propagation delays and temperature mismatch	. 89
	4.7.5	Leak analysis	91
	4.7.6	Countermeasures to the transient attack	. 93
4.8	Rema	rks on the proposed KKD proof of security	94
	4.8.1	Parametrisation of the design	. 95
	4.8.2	Continuity argument	. 96
4.9	Concl	usion	. 97
	4.9.1	Original contributions	. 98
Chapter	r 5. Tri	ustworthy Randomness for Identity Management	101

Contents

5.1	Public-	key distribution: the <i>status quo</i>	102
	5.1.1	The public-key infrastructure	102
	5.1.2	PKI failure modes	103
	5.1.3	The Web of Trust	105
	5.1.4	Identity-based cryptography	106
5.2	Anonyi	mous Auditing	106
	5.2.1	Motivation	108
	5.2.2	Related work	108
	5.2.3	Verification protocol	111
	5.2.4	Security Analysis	112
	5.2.5	Anonymisation methods	124
	5.2.6	Discussion	130
	5.2.7	Implementation	132
5.3	Distrib	uted certificate issuance	134
	5.3.1	Preliminaries	137
	5.3.2	Random verification	138
	5.3.3	Success over time in gaining false certificates	141
	5.3.4	Avoidance of repeated requests	142
	5.3.5	Implementation	144
5.4	Conclu	sion	146
	5.4.1	Original contributions	146
			147
Chapter	6. Cor	iclusions and ruture Directions	14/
6.1	Conclu	sions and contributions	148
6.2	Future	work	149
	6.2.1	Failure modes of stochastic systems	150
	6.2.2	Nonlinear sensing	150
	6.2.3	Noise-based communications	151
	6.2.4	Stochastic approaches to identity management	151

A.1	The ha	rdware platform
A.2	Theory	of operation
A.3	Design	155
	A.3.1	Analogue frontend
	A.3.2	ADC interface
	A.3.3	DSP Framework
	A.3.4	DSP
	A.3.5	Communications
	A.3.6	Noise generation
A.4	Operat	ion
	A.4.1	KKD experiment operation
	A.4.2	Testing
A.5	SCPI c	ommand reference
	A.5.1	General commands
	A.5.2	:MEASure commands
	A.5.3	:SENSe commands
	A.5.4	:OUTPut commands
A.6	Schem	atics
A.7	Softwa	re build environment
Append	ix B. So	ource Code 173
B.1	Nonlin	ear sensing
	B.1.1	Floating-point implementation
B.2	Noise-	based Communications
	B.2.1	Directional coupler
	B.2.2	Round-trip-time measurement engine
	B.2.3	Round-trip-time key establishment system
Append	ix C. T	he Allison Mixture 201
C.1	Linear	statistics of the Allison Mixture
	C.1.1	The Allison mixture

	C.1.2	Numerical results	206
C.2	Inform	nation-theoretic analysis of the Allison Mixture	207
	C.2.1	The Allison mixture	207
	C.2.2	Autoinformation of the Allison mixture sampling process	209
	C.2.3	Open questions	211
Bibliography		:	213
Acronyn	ns	2	227
Index		2	229
Biograp	hy	2	233

Abstract

No measurement system is perfect, and two varieties of error compete to frustrate their designers and operators. Random errors produce measurement-to-measurement to variation, while systematic errors result in consistently-incorrect results.

The interplay between these two phenomena has been the subject of research for many years, particularly within the area of *stochastic resonance*, which focusses upon cases where the signal-to-noise ratio of a nonlinear system can increase with the addition of noise to its input signal. While it has been demonstrated many times that noise can overcome systematic deficiencies in a measurement system, there remain open questions on how to take advantage of this in practical systems, what information can be extracted, and whether such 'randomised' systems are useful in other settings.

In this thesis, we consider this general theme in the context of two main settings: the adversarial, and the nonadversarial. In both cases, there is a significant advantage to be gained from the use of techniques that are adapted to the problem domain, in contrast to previous ad-hoc approaches that have failed to take advantage of the structures of the problems at hand.

The first part of this thesis considers the elimination of static nonlinearity from noisy measurements. We start with the phenomenon of 'classical' stochastic resonance, showing how input noise can be used to linearise the response of a nonlinear system. This phenomenon has been observed in the past, however we demonstrate that the use of nonlinear signal processing allows the linearisation to take place with far smaller levels of noise. We then investigate several approaches to the implementation of this technique, with the aim of supporting real-time operation in embedded systems and vLSI.

The remainder of the thesis concerns the use of randomness in measurements made as part of adversarial systems. This can be split into two situations: that where the operation of a system requires that measurement be difficult, and that where measurement must be straightforward. We first discuss the Kish key distribution system, a proposed classical alternative to quantum key distribution. This system claims to derive its security from the second law of thermodynamics, however these claims have been the subject of controversy. We examine the claims in detail, and show that the use of random signals does not render implausible the measurement of the system state. Finally, we describe a number of approaches to the topical problems of key distribution and identity verification. We show how various forms of multi-path probing can be treated as a form of random sampling; much like in the first section, this randomness allows for the characterisation of systematic errors, in this case the consistent changes introduced by an attacker. We then compute bounds on the probability that an attacker achieves a deception against a user taking part in this sampling process.

The first approach that we consider uses an anonymising system such as Tor or a mixnet; if all users make anonymous requests to a service in lock-step, then a malicious service cannot guarantee a self-consistent set of responses to anyone without providing the malicious response to all users. This allows the development of a statistically guaranteed consensus, and thus permits auditors to assure themselves that they have examined the same data as has been provided to other users. This provides an attractive alternative to blockchain technology, avoiding the complexity of the proof-of-work and proof-of-stake-based systems that dominate the landscape today.

We have developed a second approach that allows the random-sampling approach to be used with the existing public-key infrastructure. By demonstrating that the entities chosen to carry out the verification of an identity holder are selected at random from a substantial number of independent entities, relying parties can be confident that small numbers of compromised verifiers cannot unilaterally issue certificates for identities that they do not hold. This provides a basis for the development of highly robust distributed certificate issuance systems that do not share the current 'weakest-link' nature of the existing public-key infrastructure.

Ultimately, these systems all hold in common the use of randomness in their measurement conditions in order to characterise systematic effects. While this phenomenon has been acknowledged, its potential to characterise real systems has until now not been realised. We demonstrate that randomness, whether natural and unavoidable or artificially introduced, can ironically render far more predictable the behaviour of many systems, and in more realistic situations than have been seen in the literature to date.

Statement of Originality

I certify that this work contains no material which has been accepted for the award of any other degree or diploma in my name, in any university or other tertiary institution and, to the best of my knowledge and belief, contains no material previously published or written by another person, except where due reference has been made in the text. In addition, I certify that no part of this work will, in the future, be used in a submission in my name, for any other degree or diploma in any university or other tertiary institution without the prior approval of the University of Adelaide and where applicable, any partner institution responsible for the joint-award of this degree.

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I acknowledge the support I have received for my research through the provision of an Australian Government Research Training Program Scholarship, and an Endeavour Research Fellowship.

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Date

Signed

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Conventions

Acronyms are given in small caps—e.g. ADC, PDF—where this does not create ambiguity. Exceptions generally involve plurals, e.g. ADC/ADCs.

The main text is typeset in *Minion Pro*, with the mathematics set in *T_EX Gyre Pagella*. Figure captions and other sans-serif text are set in *Charlotte Sans*.

Publications

Papers marked ► *are directly relevant to this thesis.*

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- 1. ► L. J. Gunn, A. Allison, and D. Abbott (2013). Identification of static distortion by noise measurement. *Electronics Letters* 49(21), pp. 1321–1323. DOI: 10.1049/el.2013.2547
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- 4. ► L. J. Gunn, A. Allison, and D. Abbott (2015a). "Real-time compensation of static distortion by measurement of differential noise gain". *Proc. IEEE Workshop on Signal Processing Systems*. Belfast, United Kingdom. DOI: 10.1109/SiPS.2014.6986079
- 5. ► L. J. Gunn, F. Chapeau-Blondeau, A. Allison, and D. Abbott (2016c). Towards an information-theoretic model of the allison mixture stochastic process. *Journal of Statistical Mechanics: Theory and Experiment* 2016(5). DOI: 10.1088/1742-5468/2016/05/054041
- 6. ► L. J. Gunn, A. Allison, and D. Abbott (2017). "Safety in numbers: anonymization makes keyservers trustworthy". 10th Workshop on Hot Topics in Privacy Enhancing Technologies. Minneapolis, USA¹

¹See Gunn et al. (2016a) for the full paper.

List of Figures

1.1	Measurement bases for BB84. One of the four basis vectors is selected at	
	random, and sent to the recipient, Bob, in the form of a single polarised	
	photon. If the photon is measured with the wrong measurement basis, the	
	measurement will be wrong 50% of the time	7

- 2.1 Prior and posterior distributions of the heads-probability of a biased coin. We suppose a prior distribution of $H \sim \mathcal{N}(0.5, 0.05)$. Because the tails of the Gaussian distribution are not heavy, the posterior distribution moves only very slowly away from the unbiased value of 0.5.
- 2.3 Probability of guilt given varying numbers of unanimous line-up identifications, assuming a 50% prior probability of guilt and identification accuracies given by Foster et al. (1994). Of note is that for the case that we have plotted here where the witnesses are unanimous, with a failure rate $p_c = 0.01$ it is impossible to reach 95% certainty in the guilt of the suspect, no matter how many witnesses have been found.

2.4	Probability of guilt as a function of judges in agreement out of 23—the number used by the Sanhedrin for most capital crimes—for various contamination rates p_c . We assume as before that half of defendants are guilty, and use the estimated false-positive and false-negative rates of juries from Spencer (2007, model (2)), 0.14 and 0.25 respectively. We arbitrarily assume that a 'contaminated' trial will result in the a positive vote 95% of the time. The panel of judges numbers 23, with conviction requiring a majority of two and at least one dissenting opinion (Epstein 1961, Sanhedrin), thus requiring 13 to 22 votes inclusive in order to secure a conviction, as shown in the graph.	23
2.5	A function that tests for primality by attempting to factorise its input by brute force.	25
2.6	The acceptance rate as a function of time in memory and the number of Rabin-Miller iterations under the single-error fault model described in this thesis. An acceptance rate of 2^{-128} is normally chosen, however without error correction this cannot be achieved. The false-acceptance rate after k iterations is given by $p_{\text{fa}}[k] = 4^{-k}(1 - p_f) + p_f$, where p_f is the probability that a fault has occurred that causes a false acceptance 100% of the time. We estimate p_f to be equal to $10^{-19}T$, where T is the length of time in seconds	
	that the code has been in memory	27
3.1	Experimentally-noise-estimated amplifier voltage transfer function	36
3.2	THD of amplified signal before and after naïvely-implemented nonlinearcompensation	37
3.3	Static error of the amplifier output before and after compensation, with quantisation noise included	38
3.4	Static error of the amplifier output before and after compensation, with quantisation noise removed	38
3.5	Basis functions used for differential gain approximation	39
3.6	Block diagram of the adaptive real-time nonlinear compensator	40
3.7	Distortion compensator experimental setup	42
3.8	Real-time compensation of a distorted triangle wave	42
3.9	INL and DNL of amplifier before and after real-time compensation	44
3.10	Amplifier THD before and after compensation	45

3.11	Distorted sinusoid before and after real-time compensation	45
3.12	Compensating function construction for high-speed evaluation	47
3.13	Feedback-based algorithm for nonlinear compensator	48
3.14	Test setup for feedback-based nonlinear compensator	50
3.15	THDs before and after feedback-based compensation of distorted sinusoid .	51

4.1	The basic Diffie-Hellman key-estabishment protocol. In this diagram, $\stackrel{\$}{\leftarrow}$	
	denotes the selection of a random element from the set to the right, (G, \cdot) is	
	a finite cyclic group of order p , and g a generator for the group	57
4.2	Timestamping events for round-trip-time measurements	60
4.3	Round-trip time distribution	61
4.4	Server locations in round-trip-time measurement system	64
4.5	Bit-error-rates of timing-based protocol after information reconciliation	65
4.6	The KKD system under analysis	66
4.7	The four possible resistor states. Each time the protocol is run, the two switches are set at random, placing the system into one of the four states shown; at the bottom of each square is the mean-square line voltage for $R_a = 1\Omega$, $R_b = 2\Omega$, and $4kTB = 2$; this is only for illustrative purposes, and in practice the resistors will be of the order of several kilo-ohms. Two of the states are indistinguishable by an eavesdropper measuring only $\langle V^2 \rangle$, while Alice and Bob, who know their own selected resistor values, and so which row and column respectively the true state is in, can distinguish all four states. When running the protocol, Alice and Bob simply agree to drop any insecure bits from the generated random key	67
4.8	Directional measurement analog frontend	69
4.9	DSP block diagram of directional wave measurement device	71
4.10	An <i>s</i> -parameter model of the ккр system.	72
4.11	Log likelihood-ratio test statistics for the KKD attack measurements	76
4.12	Simulated eavesdropper error rates for the KKD system with attenuation	77
4.13	Directional coupler as constructed	78
4.14	Coupler apparatus frequency response	79
4.15	Measured eavesdropper BER for attenuation attack	80

4.16	KKD schematic with component values	86
4.17	Secrecy rate of simulated KKD system with finite line resistance	88
4.18	Secrecy rate of ideal KKD system with voltage mismatch	89
4.19	Apparent kkd noise temperatures as a function of time	90
4.20	Mismatch of apparent temperatures in KKD system at startup	91
4.21	Error rates of KKD resistor estimation via apparent temperature mismatch .	93
4.22	KKD secrecy rate with an eavesdropper using the transient attack	94
4.23	A resistive circuit containing two secret DC voltage sources V_1 and V_2 , each with a series resistance of 1 Ω . An eavesdropper can measure the voltage at two points on the line, yielding voltages V_x and V_y , which determine V_1 and V_2 if and only if $R \neq 0$.	96

5.3	A model of an anonymously-accessed service, where $\mathcal A$ is the potentially-
	malicious service, and \pounds is a leakage function that captures the information
	leaked to the adversary. In the case of Tor, for example, \pounds is the user-to-
	request mapping $R_{\rm I}$ with its domain restricted to users whose entry guards
	are surveilled by the attacker. The service accepts a set of users, and selects a
	random mapping from users to request identifiers. The adversary is given
	system-dependent partial information on the source of each request, and
	invited to provide a response to each request
5.4	Security experiment for sender-anonymity. An anonymity system, defined
	by its leakage function \pounds , is used to make requests to an adversary who aims
	provide particular messages to particular users. The adversary is asked to
	determine the users to whom each of its responses were sent; it wins if it
	correctly identifies all of the recipients
5.5	Connecting to a public keyserver via Tor and via a mix-net. The user ran-
	domly selects several relays, then adds a layer of encryption for each relay.
	After receiving a message, the relays strip their layer of encryption, revealing
	the address of the next relay. Eventually, the message reaches an <i>exit node</i> ,
	which passes it to the open internet. Anyone can contribute nodes to the
	network—including adversaries—however as the routing path is selected
	by the user, an attacker cannot gain access to the encrypted messages with
	probability better than chance. Mix-nets are composed of a chain of mixes,
	which take batches of messages, remove a layer of encryption, shuffle the
	messages, then pass them to a new mix. If at least one mix in the chain
	is honest, then an attacker cannot connect messages to their senders with
	probability better than chance
5.6	Waiting-time necessary to achieve various levels of security. We show the
	hypothesized Certificate Transparency system modelled on the Chrome auto-
	update mechanism (top), our proposed keyserver-auditing system (middle),
	and our conception of how a keyserver built on something like Continuous
	Identity and Key Management System (CONIKS) might look (bottom). We
	see that very small probabilities of equivocation are achieved within only a
	few hours, such that deanonymisation and endpoint compromise quickly
	become far more likely than chance success by a malicious service 133
5.7	Certificate issuance protocol
5.8	Probability of selecting colluding verifiers at random

5.9	Probability of obtaining a false certificate with imperfect verifiers 141
5.10	Architecture of distributed CA prototype

C.1	The Markov chain defining the sampling process S_t of the Allison mixture.
	It is parametrised by the probabilities α_0 and α_1 of leaving states 0 and 1
	respectively. When in state one, its value is equal to that of the first process
	U, and when in state two to that of the second V
C.2	The autocorrelation coefficient of the Allison mixture of $N(-10, 1)$ and
	$N(10,1)$ for various values of α_1 and α_2 . The thick line shows the case
	$\alpha_1 + \alpha_2 = 1$ in which the autocorrelation coefficient is zero
C.3	The Allison mixture of $N(-10, 1)$ and $N(10, 1)$ with varying parameters α_i .
	In (a) the low probability of switching causes the process to stay with its
	current input for long periods of time. The autocorrelation coefficient is large
	and positive. Conversely, in (a) the probability of switching is high, causing
	the sampling operation to flit between the two processes almost every cycle.
	The autocovariance is large and negative
C.4	Single-step autoinformation $I_{SS}[1]$ of the Allison mixture sampling process
	S_t as a function of the transition probabilities α_0 and α_1 , calculated according
	to Equation C.47. Note the lines of zero autoinformation along $\alpha_0 = 0$,
	$\alpha_1 = 0$, and $\alpha_0 + \alpha_1 = 1$
C.5	The exponentially-decaying autoinformation $I_{SS}[k]$ and autocovariance $R_{SS}[k]$
	of an Allison mixture sampling process with $\alpha_0 = 0.1$, $\alpha_1 = 0.1$. The slope of
	the autoinformation line is approximately double that of the autocorrelation
	line; the results of Chapeau-Blondeau (2007) hint that this may be exactly so
	asymptotically

List of Tables

2.1	The model parameters for the case of the pot for use in (2.3) with a contamina-	
	tion rate $p_c = 10^{-2}$. The <i>a priori</i> distribution of the origin is identically 50%	
	for both Britain and Italy, whether or not the pot's manufacturing process	
	has contaminated the results. As a result, the two columns of $P[F, H_i]$ are	
	identical. The columns of the measurement distribution, shown right, differ	
	from one another, thereby giving the test discriminatory power. When the	
	pot has been contaminated, the probability of a positive result is identical for	
	both samples, rendering the test ineffective	18
2.2	The model parameters for the hypothetical identity parade. In a similar	
	fashion to the first example, we assume <i>a priori</i> a 50% probability of guilt. In	
	this case, the measurement distributions are substantially assymmetric with	
	respect to innocence and guilt, unlike Table 2.1.	21
2.3	The model parameters for the Sanhedrin trial. Again, we assume an <i>a priori</i>	
	50% probability of guilt. However, the measurement distributions are the	
	results of Spencer (2007, model (2)) for juries; in contrast to the case of the	
	identity parade, the false negative rate is far lower. Despite the trial being	
	conducted by judges, we choose to use the jury results, as the judges tendancy	
	towards conviction is not reflected in the highly risk-averse rabbinic legal	
	tradition	23
2.4	Model parameters for the Rabin-Miller test on random 2000-bit numbers.	
	However, we have no choice but to assume the lower bound on the composite-	
	number rejection rate, and so this model is inappropriate. Furthermore, in an	
	adversarial setting the attacker may intentionally choose a difficult-to-detect	
	composite number, rendering the prior distribution optimistic	26
4.1	Values of $1 - \Gamma^2$ for various choices of resistor. A characteristic impedance	
	of 50 Ω is assumed.	92
5.1	Costs and security of the proposed protocol for literal-data and Merkle Tree	
	systems	131