

Eidgenössische Technische Hochschule Zürich Swiss Federal Institute of Technology Zurich

## Signal and Information Processing Laboratory (ISI)

# Annual Report 2018

Signal and Information Processing Laboratory ETH Zurich, Sternwartstr. 7, CH-8092 Zurich www.isi.ee.ethz.ch

# Foreword

by Amos Lapidoth

In the shadow of stupendous world events such as Prince Harry and Meghan Markle's wedding, we at ISI have continued to teach, mentor, supervise, and conduct research. While not quite as visible as the Royal Wedding, our efforts did not go unnoticed: a fair number of our research results were published in various journals and conference proceedings. Also noteworthy was the International Zurich Seminar (IZS'18) co-chaired by Stefan Moser and myself with tremendous support from Hans-Andrea Loeliger, Silvia Tempel, Paddy Strebel, and our friends in the Communication Technology Laboratory, in February 2018.

But all work and no play makes Jack a dull boy (and Jill a dull girl). We therefore participated in a fantastic hiking day to Stoos organized by Paddy Strebel. There we checked out the world's steepest cable railway. (We are, after all, engineers, and we get excited about such things.) The walk from Klingenstock to Fronalpstock offered amazing views, and the coffee that Stefan Moser prepared for us hit the spot.

The brave (crazy?) among us also enjoyed (much as one enjoys a terrifying roller coaster) a ski day at Engelberg, with chair lifts swinging uncontrollably from side to side. (No one will be surprised to learn that this excursion was also organized by Paddy Strebel.) Fortunately all returned happy and healthy.

Finally, we have also had a change of personnel in 2018, with Tibor Keresztfalvi leaving us (with a Ph.D. in hand) and with Raphael Keusch embarking on one. We wish him and all our students the best of luck, both at ETH and in the mountains.

# **Contents**



# **1 People**



**Secretaries:** Rita Hildebrand Silvia Tempel

# **2 Teaching**

### **2.1 Regular Courses**

- *Discrete-time and Statistical Signal Processing*, Prof. Loeliger (Bachelor & Master)
- *Communication and Detection Theory*, Prof. Lapidoth (Bachelor)
- *Information Theory I*, Prof. Lapidoth (Master)
- *Information Theory II*, Prof. Lapidoth (Master)
- *Signal Analysis, Models and Machine Learning*, Prof. Loeliger (Master)
- *Algebra and Error Correcting Codes*, Prof. Loeliger (Master)

#### **Courses by external Lecturers**

- *Acoustics I*, Dr. Kurt Heutschi (Master)
- *Acoustics II*, Dr. Kurt Heutschi (Master)
- *Analog Signal Processing and Filtering*, Dr. Hanspeter Schmid (Master)

#### **2.2 Lab Courses**

- Fachpraktika*,* Federico Wadehn
- *Blackfin DSP,* Boxiao Ma
- *Electronic Circuits and Signals Exploration Laboratory,* Hampus Malmberg

### **2.3. Student Projects**



#### **Semester Projects, Fall Term 2018**



#### **Bachelor Group Project, Spring Term 2018**



#### **Master Projects, Spring Term 2018**



#### **Master Projects, Fall Term 2018**



# **3 Research**

### **3.1 General Research Areas**

#### **Information Theory and Error Correcting Codes**

- Multi-user information theory
- Network Coding
- Combined source-channel coding
- Multi-path channels and fading channels
- Optical channels
- Error correcting codes

#### **Signal Processing**

- Fundamentals and applications of factor graphs
- State-space methods
- Sparsity and unsupervised signal decomposition
- Imaging and tomography
- "Neural" computation and signal processing
- Analog-to-digital conversion

### **3.2 Current Research Topics with Prof. Lapidoth**

#### **Guessing with Distributed Encoders**

Two correlated sources emit a pair of sequences, each of which is observed by a different encoder. Each encoder produces a rate-limited description of the sequence it observes, and the two descriptions are presented to a guessing device that repeatedly produces sequence pairs until correct. The number of guesses until correct is random, and it is required that it have a moment (of some prespecified order) that tends to one as the length of the sequences tends to infinity. The description rate pairs that allow this are characterized in terms of the Renyi entropy and the Arimoto-Renyi conditional entropy of the joint law of the sources. This solves the guessing analog of the Slepian-Wolf distributed source-coding problem.

Applications to the distributed storage of passwords are examined.

#### **Multiplexing Zero-Error and Rare-Error Communications over a Noisy Channel**

Two independent data streams are to be transmitted over a noisy discrete memoryless channel with noiseless (ideal) feedback. Errors are tolerated only in the second stream, provided that they occur with vanishing probability. The rate of the error-free stream cannot, of course, exceed the channel's zero-error feedback capacity, and nor can the sum of the streams' rates exceed the channel's Shannon capacity. Using a suitable coding scheme, these necessary conditions are shown to characterize all the achievable rate pairs. Planning for the worst channel behavior-as is needed to achieve zero-error communication-and planning for the typical channel behavior-as is needed to communicate near the Shannon limit-are thus not incompatible.

It is further shown that feedback may be beneficial for the multiplexing problem even on channels on which it does not increase the zero-error capacity.

#### **The Zero-Error Feedback Capacity of State-Dependent Channel**

One of the longest-standing open problems in Information Theory is to compute the zero-error capacity of a general discrete memoryless channel, i.e., the highest rate at which error-free communication is possible over the channel. Perhaps surprisingly, in the presence of feedback, this problem was solved by Shannon some 60 years ago. Unlike many other problems in Information Theory, the zero-error capacity becomes simpler in the presence of feedback. Motivated by this observation, we study the zero-error feedback capacity of state-dependent channels when the state information is revealed to the encoder either strictly-causally, causally, or non causally.

#### **Correlated sources over a Noisy Mutiple-Access Channel**

On the multiple-access channel (MAC), which models many-to-one communications, the sourcechannel separation does not always hold: it need not be optimal to describe the source sequences using bit streams of the rates that are optimal with respect to the allowed distortion and to then send the bits on the MAC with small probability of error. It is sometimes beneficial to exploit the correlation between the sources in order to build correlation between the transmitted symbols. If not optimal, how far from optimal is source-channel separation? To answer this question, we need lower bounds on the achiev distortions that hold transmission schemes and hence also for the optimal scheme. Finding such bounds is the aim of this project.

#### **The Rate- and-State Capacity**

The Rate-and-State capacity of a state-dependent channel with a state-cognizant encoder is the highest possible rate of communication over the channel when the decoder - in addition to reliably decthe data - must also reconstruct the state sequence with some required fidelity. Here we calculate this capacity in the presence of output-feedback, when the state reconstruction fidelity is measured using a single-letter distortion function and the state sequence is revealed to the encoder in one of two different ways: strictly-causally or causally.

#### **Mismatched Decoding in the Presence of Feedback**

For a given channel and a given decoding rule, the mismatch capacity is the highest rate at which reliable communication is possible on the channel using the given decoding rule. How to compute the mismatch capacity is a long-standing open problem in Information Theory. Here we study this problem in the presence of a feedback link from the channel's output to the encoder. We show that - although feedback does not increase the Shannon capacity of memoryless channels - feedback can increase the mismatch capacity. In fact, in its presence, the mismatch capacity may equal the Shannon capacity even when the decoding rule differs significantly from the maximum-likelihood rule.

### **3.3 Current Research Topics with Prof. Loeliger**

#### **Fundamentals and Applications of Factor Graphs**

Factor graphs are a graphical notation for system models and algorithms in a large variety of fields including error correcting codes, signal processing, statistical physics, linear algebra, and more. We find factor grahs to be very helpful in most of our research work, and we continue to develop the approach.

#### **State-Space Method for Signal Analysis**

Most of our work in signal processing is based on linear state space models. Using IIR (infinite impulse response) models decouples the model order from the sampling rate and allows effortless transitions between discrete time and continuous time. We have extended such models to pulse-like (wavelet-like) signals that are localized anywhere in time. For given observations and unknown localization the corresponding model likelihood is then itself a function of time, i.e., a signal. The computation of such a likelihood signal leads to the concept of a likelihood filter (or feature detection filter), a generalization of a matched filter.

#### **Multi-window Models and Recursive Model Fitting Beyond Least Squares**

In an extension of state space methods, we have discovered the power of multi-sequence windows on the one hand and of polynomial cost functions beyond least squares.

#### **Sparsity and Unsupervised Signal Decomposition**

Normal priors with unknown variance (NUV) promote sparsity and blend well with expectation maximization (EM). For linear state space models, this approach can be used for estimating impulsive signals, detecting localized events, smoothing with occasional jumps in the state space, and for detecting and removing outliers. Combined with system identification (learning) by EM, this approach leads to very general and versatile methods for unsupervised signal and separation and decomposition. The actual computations boil down to multivariate Gaussian message passing (i.e., variations of Kalman smoothing).

#### **"Neural" Computation and Signal Processing**

Likelihood filters (features detection filters) as above can be cascaded into a new sort of neural network. A key insight here is that such networks should internally work with spikes (sparse multichannel signals) rather than with "continuous" signals. The exploration of this approach has only just begun.

#### **Imaging and Tomography**

We also use NUV priors (see above) for images (in 2 or 3 dimensions). In particular, we use this technique for tomographic image estimation. In this case, the actual computations boil down to iterative scalar Gaussian message passing.

#### **Analog Computation and Analog-to-Digital Conversion**

We have a long-standing interest in analog computation and analog circuits for information processing. Our recent research in this area has focused on analog-to-digital converters. Moreover, some of our "neural" computation algorithms (see above) are easily implementable as analog circuits.

### **3.4 Publications**





### **3.5 Completed PhD Theses**

Tibor Keresztfalvi, *Some Data are More Important than Others.* ETH Diss. 25430 (Prof. Lapidoth). Co-examiner: Prof. Osvaldo Simeone, King's College London, UK

# **4 Trips and Talks**

### **4.1 Participation in Conferences and Meetings**



## **4.2 Additional Lectures/Talks**

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## **4.3 Local Lectures and Seminars by Invited Speakers**

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## **5 Service Activities**

## **5.1 Conference Organization**

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## **5.2 Other Service Activities**

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