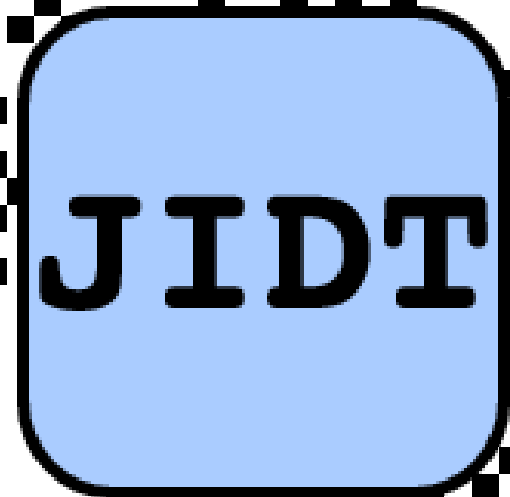


# Tutorial: Empirical analysis of information content and flows in neural data using JIDT

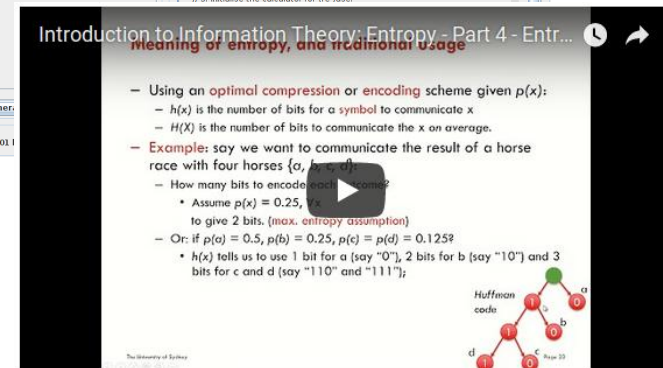
Dr. Joseph Lizier



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- Appreciation of what information theory can tell us about neural data
  - E.g. regarding dynamics of information processing
- Understanding of what JIDT offers and able to get started:
  - AutoAnalyser
  - Short course
- Primary references:
  - Lizier, "JIDT: An information-theoretic toolkit for studying the dynamics of complex systems", Frontiers in Robotics and AI, 1:11, 2014.
  - Github: <http://github.com/jlizier/jidt/>
  - Short course: <http://bit.ly/jidt-course-alpha>



# A game about information: Guess Who? (Hasbro)

## 1. Game board / rules

a. Play yourself online

## 2. What did we learn from this game?

- a. What are the best/worst questions to ask or strategies?
- b. What types of information did we encounter?



# Defining information – first pass

– JL: “*Information is all about questions and answers*”

- **Information** is the amount by which
  - one variable (an answer/signal/measurement)
  - reduces our **uncertainty** or **surprises** us
  - about another variable.

- We need to quantify both:
  - Uncertainty (**entropy**)
  - Uncertainty reduction (**information**)

– This was quantified by Claude Shannon



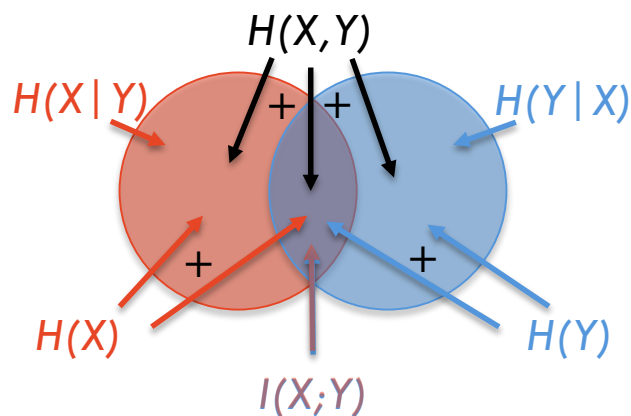
C. E. Shannon. A mathematical theory of communication. Bell System Technical Journal, 27(3–4):379–423, 623–656, 1948.

T. M. Cover and J. A. Thomas. Elements of Information Theory. Wiley-Interscience, New York, 1991.

D. J. C. MacKay. Information Theory, Inference, and Learning Algorithms. Cambridge University Press, Cambridge, 2003

# Fundamental measures

	Entropy (uncertainty)	Mutual information
Average	$H(X) = \sum_{x \in A_x} -p(x) \log_2 p(x)$	$I(X; Y) = H(X) + H(Y) - H(X, Y)$ $I(X; Y) = H(X) - H(X Y)$
Pointwise/local	$h(x) = -\log_2 p(x)$	$i(x; y) = \log_2 \frac{p(x y)}{p(x)}$



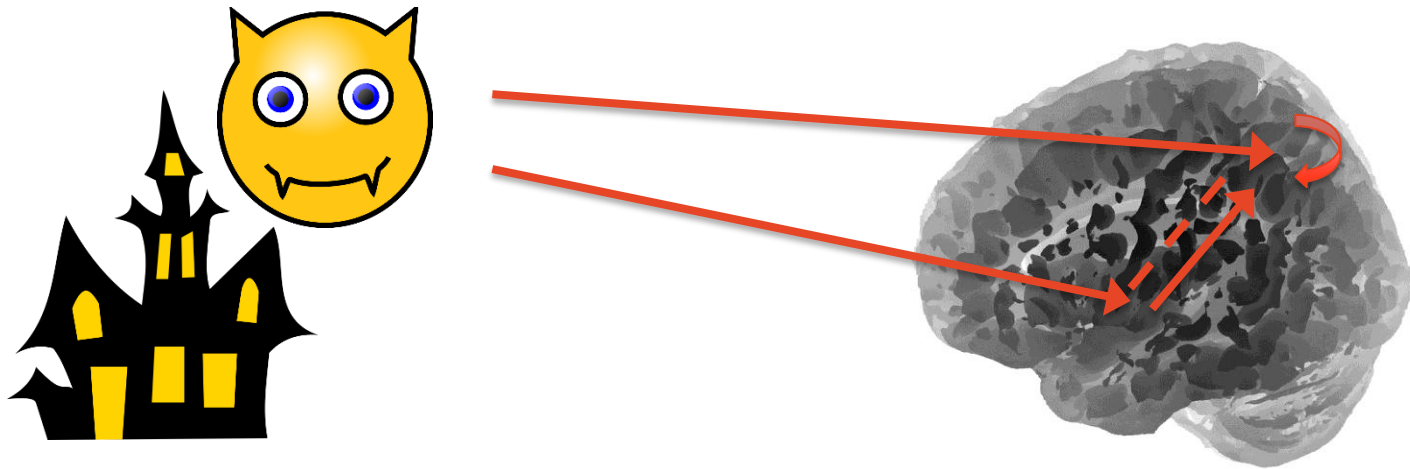
Can define:

- Joint versions,
  - e.g.  $H(X, Y)$  or  $H(\mathbf{X})$ ; and
- Conditional versions,
  - e.g.  $H(X|Y)$ ,  $I(X; Y|Z)$

Uncertainties and information in Guess Who?

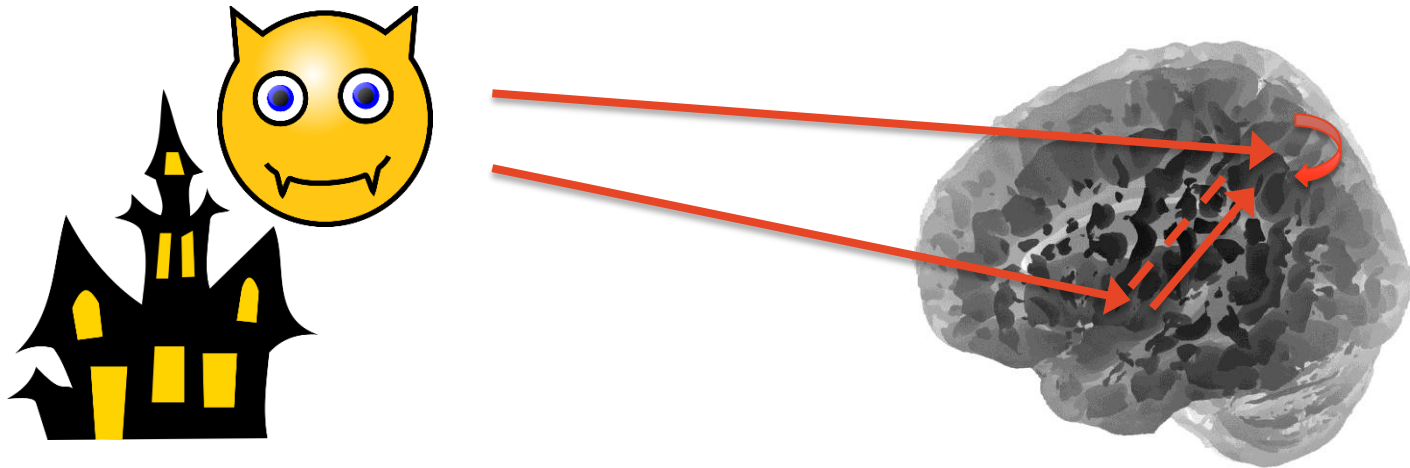
# Why use information theory for neural data analysis?

- Allows us to answer questions that are naturally phrased in this domain
  - In a model-free way
  - Captures non-linearities
  - Estimators for different data types, and multivariates
- Aligns with descriptions of dynamics and information processing

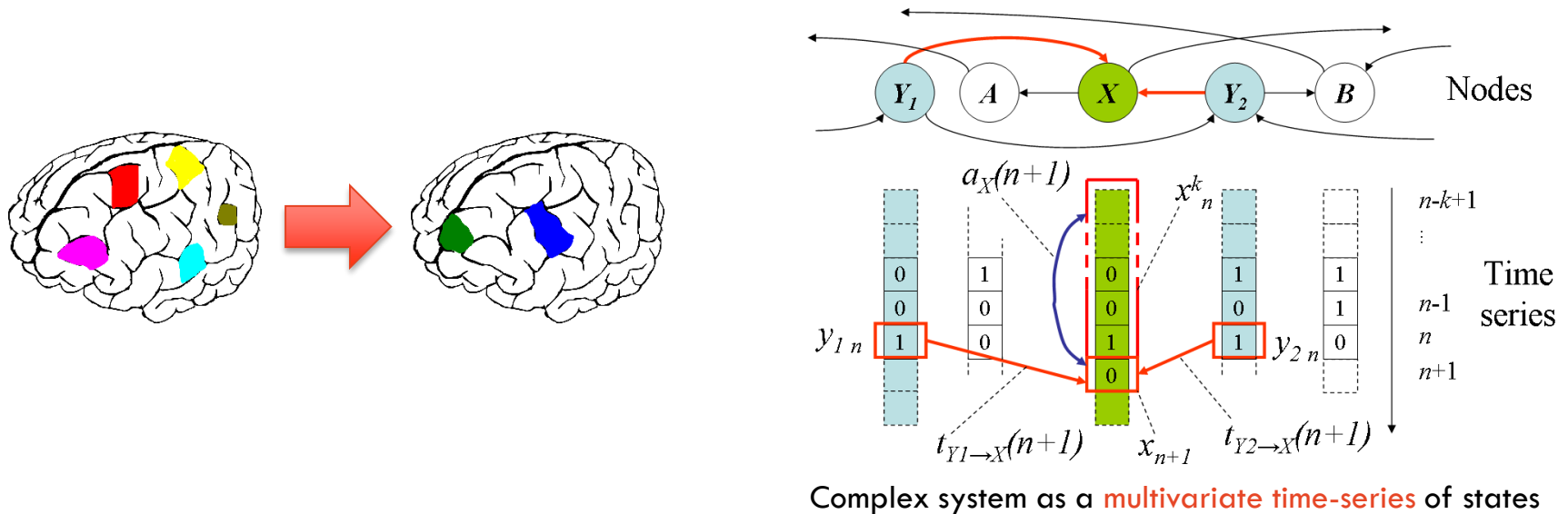


# What can we use information theory to ask?

- The nature of neural codes:
  - Where and how much uncertainty?
  - Bridge between **Marr's task and implementation levels** with ML:
    - Which response features carry information about stimulus?
    - Which **specific** responses carry information about which stimuli?
  - **Functional relationships** between neural responses?
  - **Multivariate decompositions** of information?



# What can we use information theory to ask?



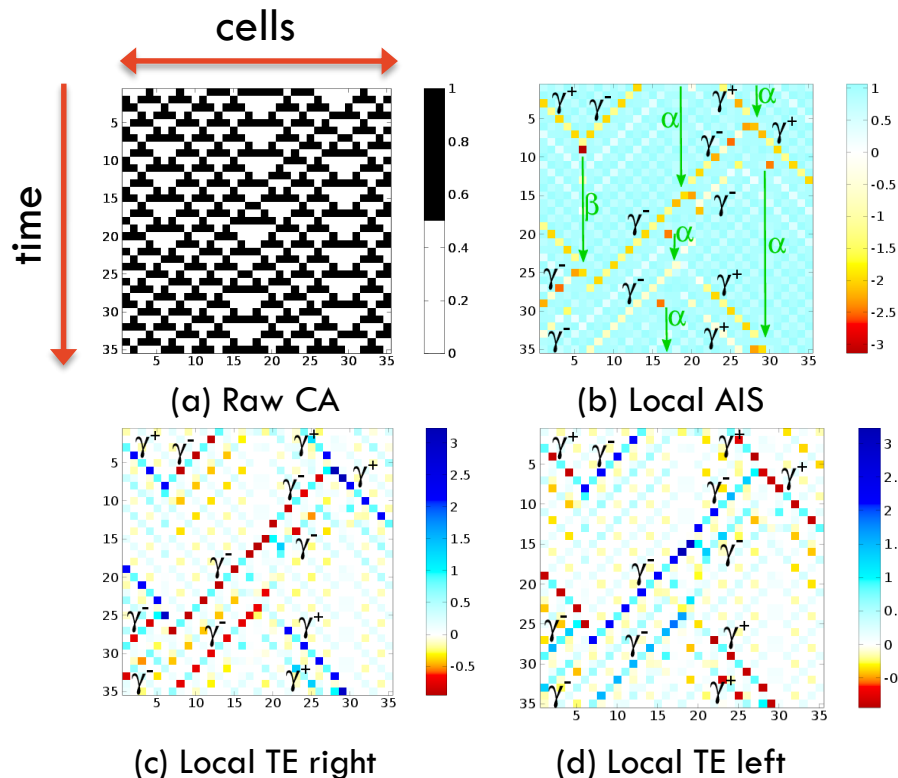
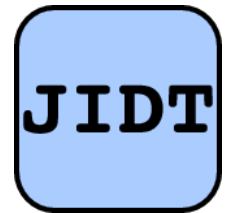
- How can we *model* neural **information processing dynamics**?
  - It is the output of a **local computation** within the system
  - Model in terms of **information storage and transfer**
  - Bridging Marr's **algorithmic and implementation levels**
  - Establishing an information-theoretic footprint

J.T. Lizier, "The local information dynamics of distributed computation in complex systems", Springer: Berlin/Heidelberg, 2013

M. Wibral, J.T. Lizier and V. Priesemann, "[Bits from Brains for Biologically-inspired Computing](#)", Frontiers in Robotics and AI, vol. 2, 5, 2015



# Example: Computational role of coherent structure in CAs



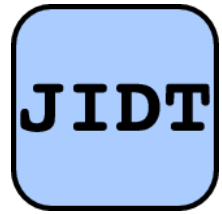
Blinkers and background domains are dominant storage entities!

Gliders are dominant transfer entities!

Links algorithmic and implementation levels

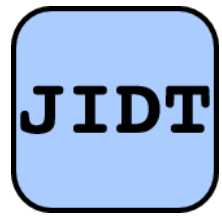
J. T. Lizier, M. Prokopenko, & A. Y. Zomaya. "Local information transfer as a spatiotemporal filter for complex systems". *Physical Review E*, 77(2):026110, 2008.  
J.T. Lizier, "JIDT: An Information-Theoretic toolkit for studying the dynamics of complex systems". *Frontiers in Robotics and AI*, 1:11, 2014.

# Java Information Dynamics Toolkit (JIDT)



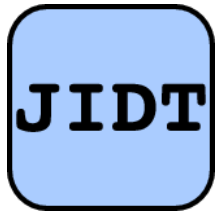
- JIDT provides a standalone, open-source (GPL v3 licensed) implementation of information-theoretic measures of information processing in complex systems
- JIDT includes implementations:
  - Principally for transfer entropy, mutual information, their conditional variants, active information storage etc;
  - For both discrete and continuous-valued data;
  - Using various types of estimators (e.g. Kraskov-Stögbauer-Grassberger, linear-Gaussian, etc.).
- Available on github: <http://github.com/jlizier/jidt/>

# Java Information Dynamics Toolkit (JIDT)



- JIDT is written in Java but directly usable in Matlab/Octave, Python, R, Julia, Clojure, etc.
- JIDT requires almost zero installation.
- JIDT is associated with:
  - A paper describing its design and usage:
    - J.T. Lizier, Frontiers in Robotics and AI 1:11, 2014; arXiv:[1408.3270](https://arxiv.org/abs/1408.3270)
  - Full Javadocs and [wiki](#);
  - A [course](#) (*in progress*; tutorial and exercises for now).
  - A suite of [demonstrations](#), including in each of the languages listed above;
  - A [GUI](#) for push-button analysis and code template generation;
- Code credits: JL, Ipek Özdemir, Pedro Martínez Mediano, ...

# Auto Analyser GUI (Code Generator)



- Computing MI could not be easier:

The screenshot shows the JIDT Mutual Information Auto-Analyser GUI. The window is titled "JIDT Mutual Information Auto-Analyser". It is divided into two main panels: "Calculation parameters" on the left and "Generated code" on the right.

**Calculation parameters panel:**

- Calculator Type:** A dropdown menu set to "Discrete".
- Data file:** A text box containing "objects/JIDT/demos/data/2coupledBinaryColsUseK2.txt" and a "Select" button below it. A note below the text box says "Valid data file with 1000 rows and 2 columns".
- All pairs?:** An unchecked checkbox.
- Source column:** A text box containing "0".
- Destination column:** A text box containing "1".
- Add stat. signif.?:** Two unchecked checkboxes labeled "analytically?".
- Properties table:** A table with two columns: "Property name" and "Property value".

Property name	Property value
base	2
time difference	0
- Buttons:** A checked checkbox "Compute result?" and a "Generate code and Compute" button.

**Generated code panel:**

- Three tabs: "Java", "Python", and "Matlab".
- A text box containing "... Awaiting new parameter selection (press compute) ...".

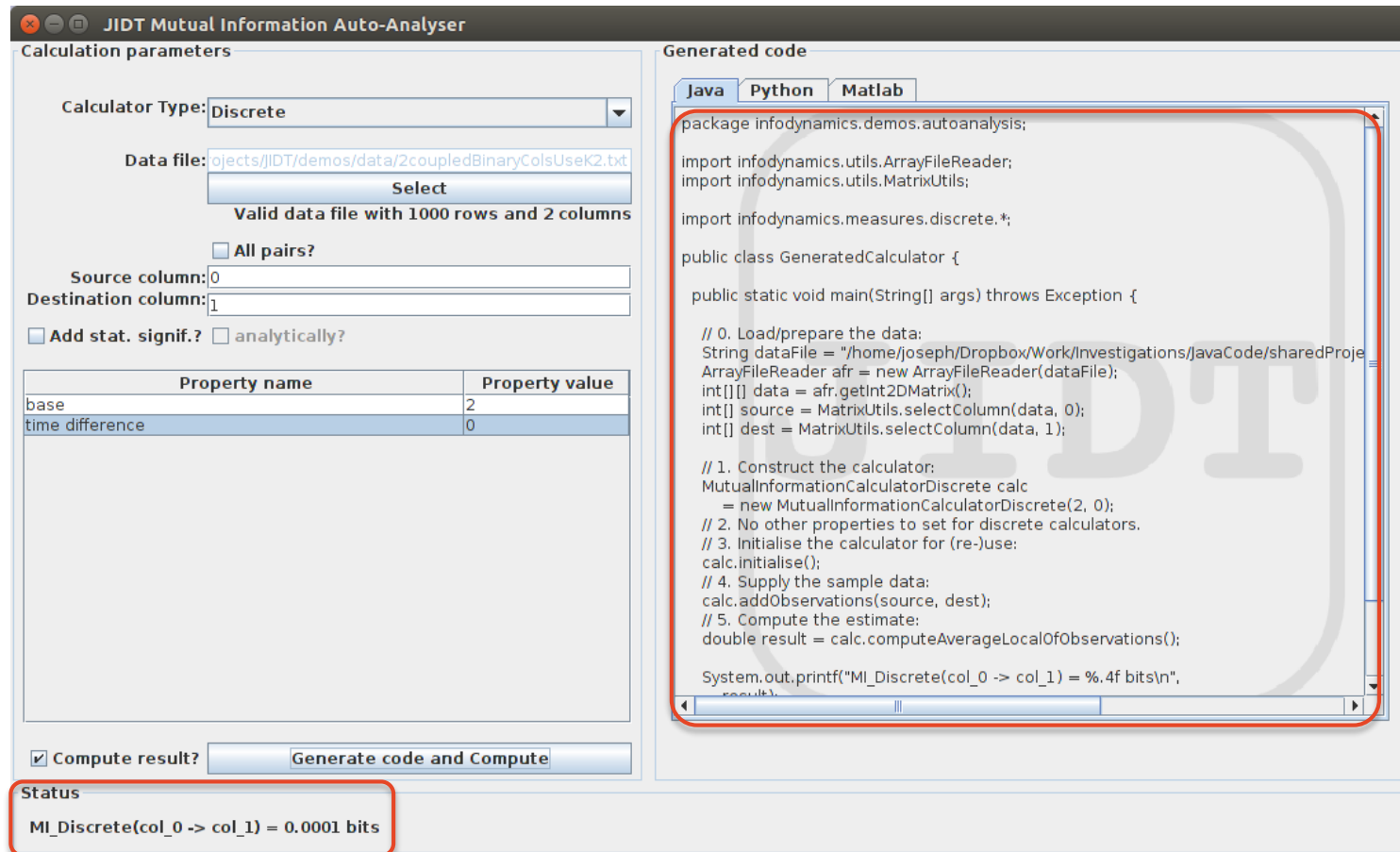
**Status panel:** A section at the bottom of the window.

Just follow the GUI:

1. Select estimator
2. Select data file
3. Identify source/target columns in data OR click "All pairs"
4. Fill out properties (use tool tip for descriptions)
5. Click "Compute"

# Auto Analyser GUI (Code Generator)

- Clicking “Compute” then gives you:
  1. The estimated result of the measure
  2. Code to generate this calculation in Java, Python and Matlab



**JIDT Mutual Information Auto-Analyser**

**Calculation parameters**

Calculator Type: **Discrete**

Data file: **objects/JIDT/demos/data/2coupledBinaryColsUseK2.txt**  
 Select  
 Valid data file with 1000 rows and 2 columns

☐ All pairs?

Source column: **0**  
 Destination column: **1**

☐ Add stat. signif. ☐ analytically?

Property name	Property value
base	2
time difference	0

☒ Compute result? **Generate code and Compute**

**Status**  
 MI\_Discrete(col\_0 -> col\_1) = 0.0001 bits

**Generated code**

**Java** **Python** **Matlab**

```
package infodynamics.demos.autoanalysis;

import infodynamics.utils.ArrayFileReader;
import infodynamics.utils.MatrixUtils;

import infodynamics.measures.discrete.*;

public class GeneratedCalculator {

    public static void main(String[] args) throws Exception {

        // 0. Load/prepare the data:
        String dataFile = "/home/joseph/Dropbox/Work/Investigations/JavaCode/sharedProje
        ArrayFileReader afr = new ArrayFileReader(dataFile);
        int[][] data = afr.getInt2DMatrix();
        int[] source = MatrixUtils.selectColumn(data, 0);
        int[] dest = MatrixUtils.selectColumn(data, 1);

        // 1. Construct the calculator:
        MutualInformationCalculatorDiscrete calc
            = new MutualInformationCalculatorDiscrete(2, 0);
        // 2. No other properties to set for discrete calculators.
        // 3. Initialise the calculator for (re-)use:
        calc.initialise();
        // 4. Supply the sample data:
        calc.addObservations(source, dest);
        // 5. Compute the estimate:
        double result = calc.computeAverageLocalOfObservations();

        System.out.printf("MI_Discrete(col_0 -> col_1) = %.4f bits\n",
            result);
    }
}
```

# Analysing neural information processing

What can it tell us about neural information processing:

1. Characterising different **regimes** of behavior
2. **Space-time dynamics** of information processing
3. **Effective network modelling**

# 1. Characterising different regimes of behaviour

**Aim:** to characterise behaviour and responses in terms of information processing;

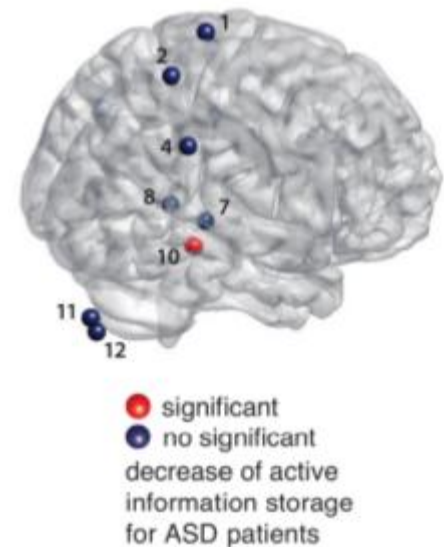
- E.g. different stimuli or neural conditions

MEG studies indicate lower resting-state information storage overall and in:

- hippocampus [1],
- precuneus, posterior cingulate cortex, supramarginal gyrus [2].

of Autism Spectrum Disorder subjects.

→ Use/precision of prior reduced



[1] C. Gómez, et al., “Reduced predictable information in brain signals in autism spectrum disorder”, *Frontiers in Neuroinformatics*, 8:9+, 2014.

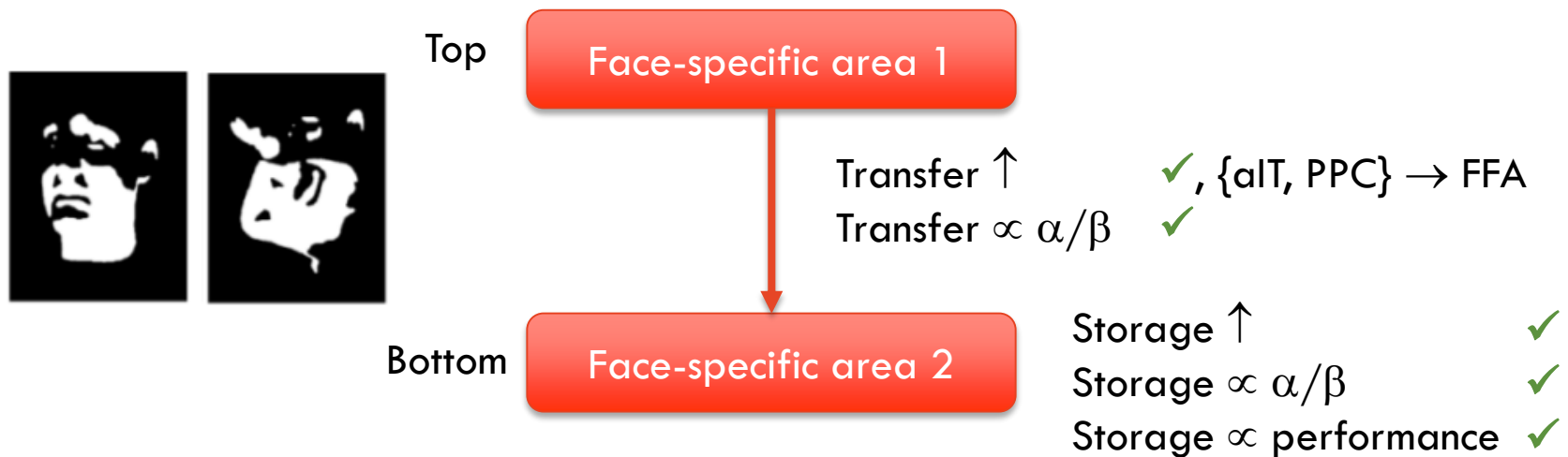
[2] A. Brodski-Guerniero, et al., “Predictable information in neural signals during resting state is reduced in autism spectrum disorder”, *Human Brain Mapping*, 39(8):3227–3240, 2018

## 2. Space-time characterization of info processing

Aim:

- Highlight info processing hot-spots locally;
- Use info processing to explain dynamics;
- Validate conjectures on neural information processing

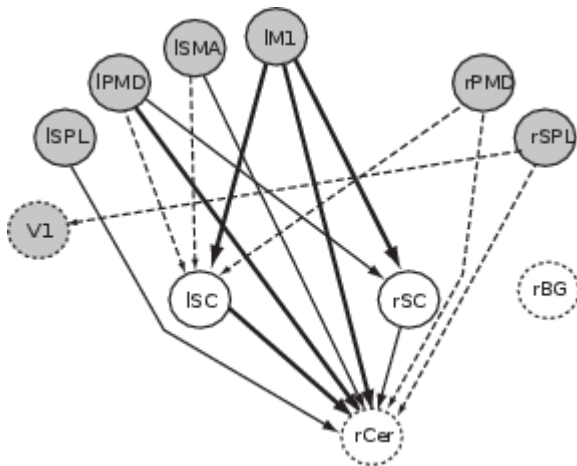
Predictive coding suggests that in a Mooney face/house detection experiment, when priming for a face:





### 3. Effective network modelling

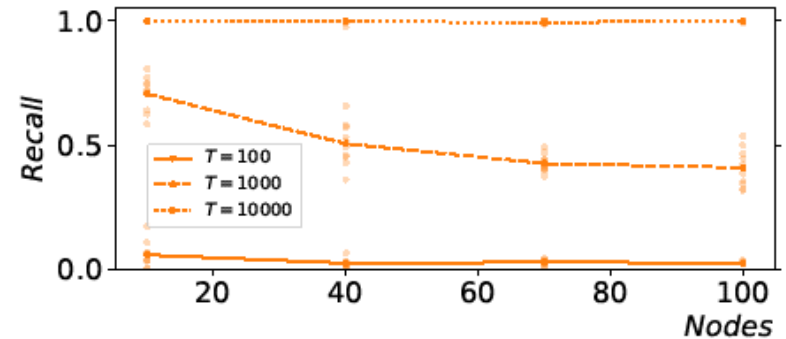
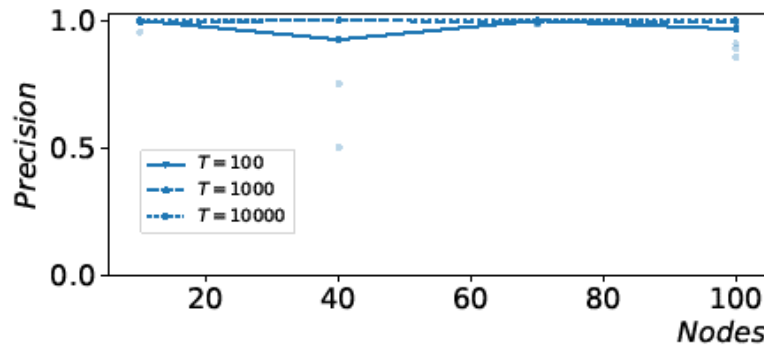
- Information transfer is ideally placed for the “*inverse problem*”
  - effective connectivity analysis – inferring a “*minimal circuit model*” that can explain observed dynamics



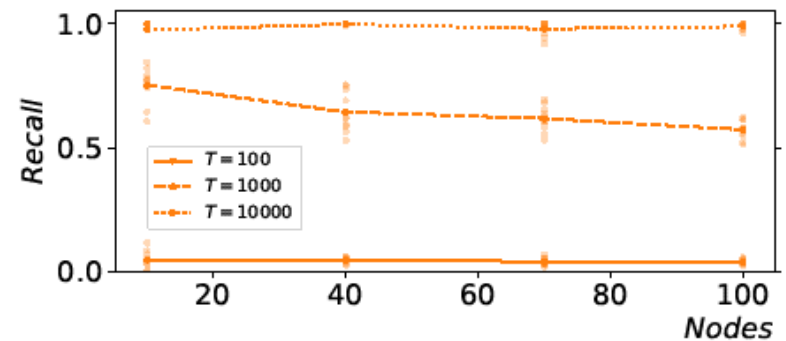
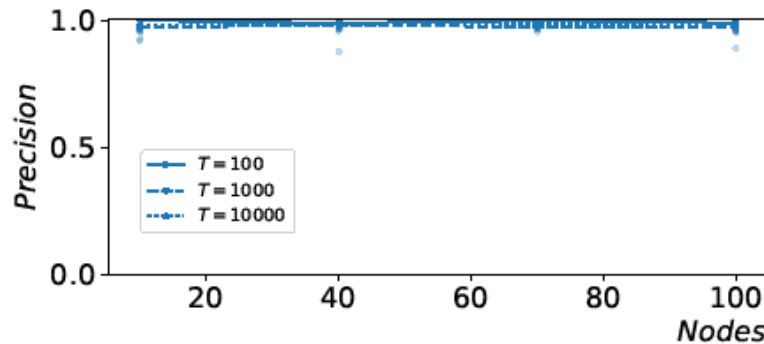
$$\begin{aligned} & \text{TRENTOOL} \\ & + \text{JIDT} \\ & + \text{Multivariate extensions, GPU \& efficiencies} \\ & = \text{https://github.com/pwollstadt/IDTxI} \end{aligned}$$

### 3. Effective network modelling

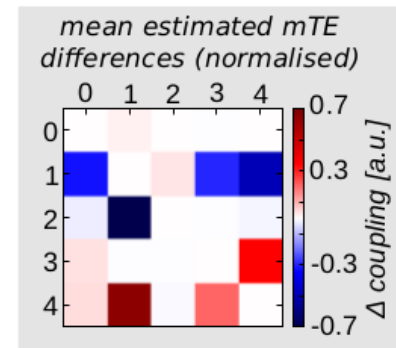
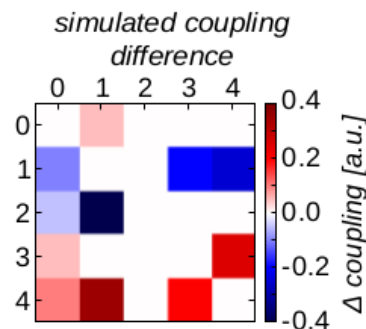
Vector auto-regressive



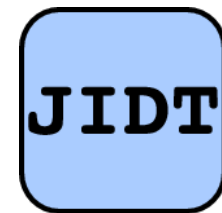
Coupled logistic maps



Group differences

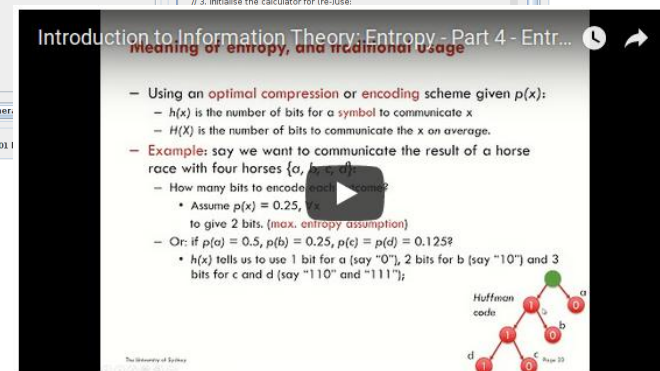
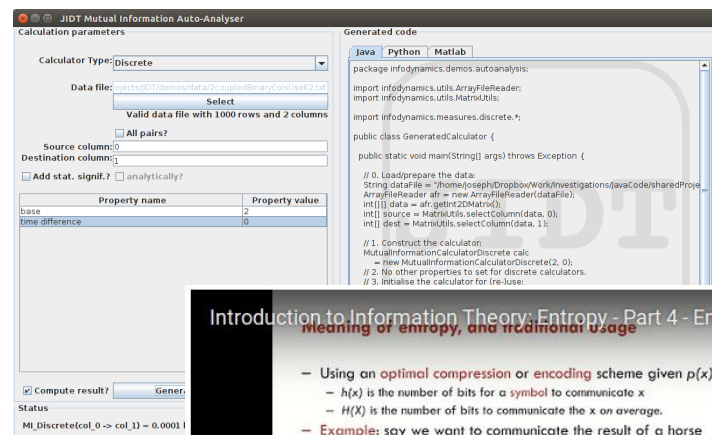


# Empirical analysis using JIDT: summary



Our session outcomes were:

- Appreciation of what information theory can tell us about neural data
  - E.g. regarding dynamics of information processing
- Understanding of what JIDT offers and able to get started:
  - AutoAnalyser
  - Short course
- Primary references:
  - Lizier, "JIDT: An information-theoretic toolkit for studying the dynamics of complex systems", Frontiers in Robotics and AI, 1:11, 2014.
  - Github: <http://github.com/jlizier/jidt/>
  - Short course: <http://bit.ly/jidt-course-alpha>



# Questions



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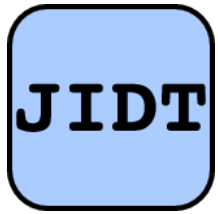
## Why implement in Java?

- Platform agnostic, requiring only a JVM;
- High performance coupled with
- Object-oriented code, with a hierarchical design to interfaces for each measure, allowing dynamic swapping of estimators for the same measure;
- JIDT can be directly called from Matlab/Octave, Python, R, Julia, Clojure, etc, adding efficiency for higher level code;
- Automatic generation of Javadocs.



# Installation

- <https://github.com/jlazier/jidt/wiki/Installation>
- **Beginners:**
  1. Download the latest full distribution by following the Download link at <https://github.com/jlazier/jidt/>
  2. Unzip it to your preferred location for the distribution
- **Advanced users:**
  1. Take a git fork/clone at <https://github.com/jlazier/jidt/>
    - a. Run `ant` (or better `ant dist`)
- To be able to use it, you will need the `infodynamics.jar` file on your classpath.
- That's it!



# Installation – caveats

1. You'll need a JRE installed (Version  $\geq 6$ )
  - Comes automatically with Matlab installation (maybe with some Octave-java or Python-JType installations)
2. Advanced users / developers, you need:
  1. full [Java SE / JDK](#) to develop in Java or to change the source code;
  2. [ant](#) if you want to rebuild the project using build.xml;
  3. [junit](#) if you want to run the unit tests.
  4. CUDA installation if you want to utilise GPU (*documentation to come*)
3. Additional preparation may be required to use JIDT in GNU Octave or Python ...

# Why use JIDT?



- JIDT is unique in the combination of features it provides:
  - Large array of measures, including all conditional/multivariate forms of the transfer entropy, and complementary measures such as active information storage.
  - Wide variety of estimator types and applicability to both discrete and continuous data



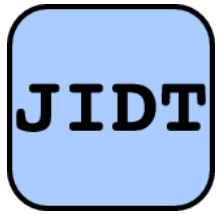
# Measure-estimator combinations

– As of version 1.2: (adapted table from paper)

Measure		Discrete estimator	Continuous estimators			
Name	Notation		Gaussian	Box-Kernel	Kraskov <i>et al.</i> (KSG)	Permutation
Entropy	$H(X)$	✓	✓	✓	*	
Entropy rate	$H_{\mu}X$	✓	<i>Use two multivariate entropy calculators</i>			
Mutual information (MI)	$I(X; Y)$	✓	✓	✓	✓	
Conditional MI	$I(X; Y   Z)$	✓	✓		✓	
Multi-information	$I(\mathbf{X})$	✓		✓ <sup>u</sup>	✓ <sup>u</sup>	
Transfer entropy (TE)	$T_{Y \rightarrow X}$	✓	✓	✓	✓	✓ <sup>u</sup>
Conditional TE	$T_{Y \rightarrow X Z}$	✓	✓ <sup>u</sup>		✓ <sup>u</sup>	
Active information storage	$A_X$	✓	✓ <sup>u</sup>	✓ <sup>u</sup>	✓ <sup>u</sup>	
Predictive information	$E_X$	✓	✓ <sup>u</sup>	✓ <sup>u</sup>	✓ <sup>u</sup>	
Separable information	$S_X$	✓				

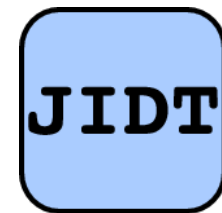
– More now, and more coming (including Partial Information Decomposition) ...

# Why use JIDT?



- JIDT is unique in the combination of features it provides:
  - Large array of measures, including all conditional/multivariate forms of the transfer entropy, and complementary measures such as active information storage.
  - Wide variety of estimator types and applicability to both discrete and continuous data
  - Local measurement for all estimators;
  - Statistical significance calculations for MI, TE;
  - No dependencies on other installations (except Java);
  - Lots of demos and information on website/wiki:
    - <https://github.com/jlazier/jidt/wiki>
  - GUI tool for easy calculation and code template generation!

# Demonstrations



- JIDT is distributed with the following demos:
  - **Auto-analyser GUI** (code generator)
  - Simple Java Demos
    - Mirrored in Matlab/Octave, Python, R, Julia, Clojure.
  - Recreation of Schreiber's original transfer entropy examples;
  - Information dynamics in Cellular Automata;
  - Detecting interaction lags;
  - Interregional coupling;
  - Behaviour of null/surrogate distributions;
  - ...
- All have documentation (PDF and wiki pages) provided to help run them.