

# Quantifying information synergy

## Using intermediate variables

Quax R, Har-Shemesh O, Sloot P. *Quantifying synergistic information using intermediate stochastic variables.*

Entropy. 2017 Feb;19(2):85.

# Counterexample?

$$X_1, X_2 \mapsto S_1$$

$$S_1 \equiv (2 - X_1 + X_2) \pmod{3}$$

$$X_1, X_2 \mapsto S_1, S_2$$

$$S_2 = (X_1 + X_2) \pmod{3}$$

		$X_2$					$X_2$		
		┌───────────┐					┌───────────┐		
{	$X_1$	2	0	1	0	1	2	0	2
		1	2	0	1	2	0	2	0
		0	1	2	2	0	1	0	1
		$S_1(X_1, X_2)$					$S_2(X_1, X_2)$		

$$I(X_1 : S_i) = I(X_2 : S_i) = 0$$

$$I(X_1 : S_1, S_2) = \log_2 3$$

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$$I_{\text{syn}}(X_1, X_2 : S_i) = \log_2 3$$

$$I(S_1 : S_2) = 0$$

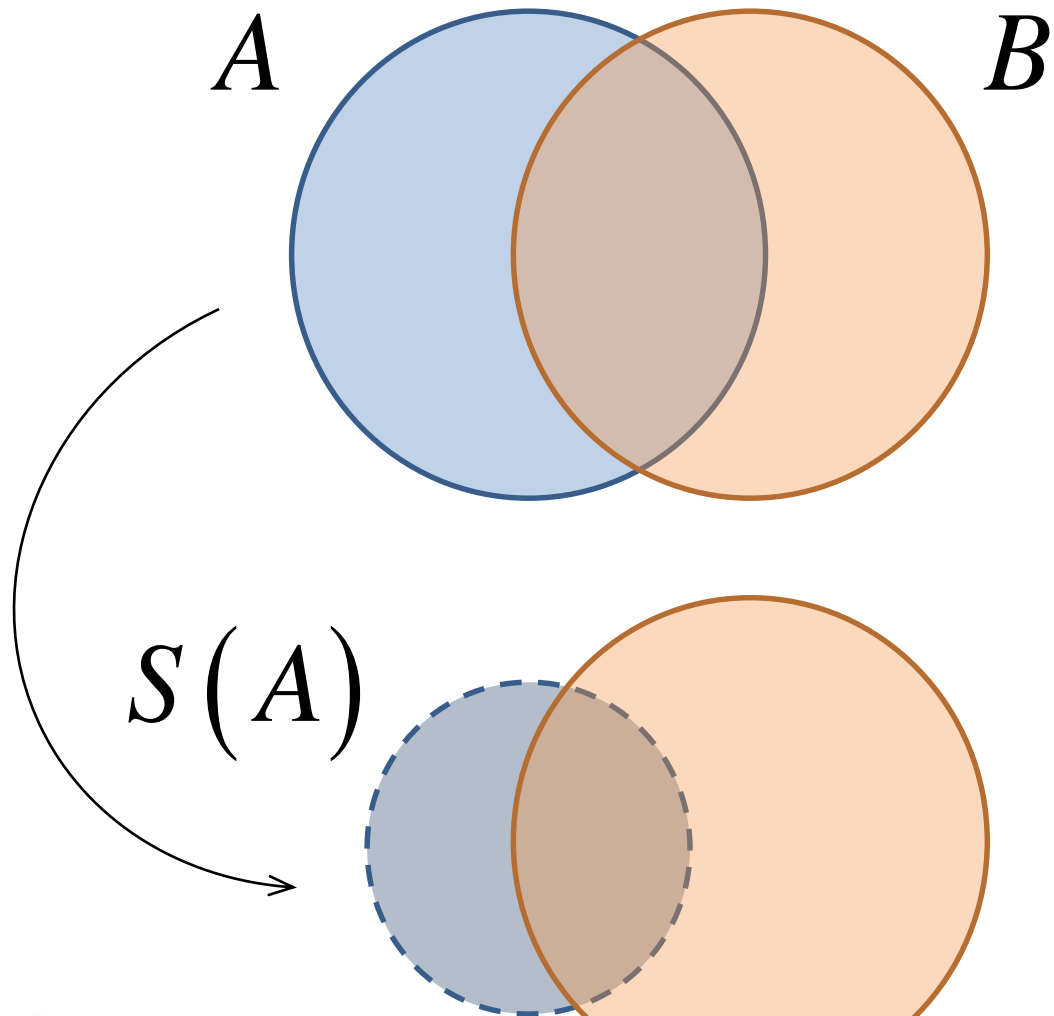
$$I_{\text{syn}}(X_1, X_2 : S_1, S_2) = 0$$

An alternative starting point for defining

# **SYNERGISTIC INFORMATION**



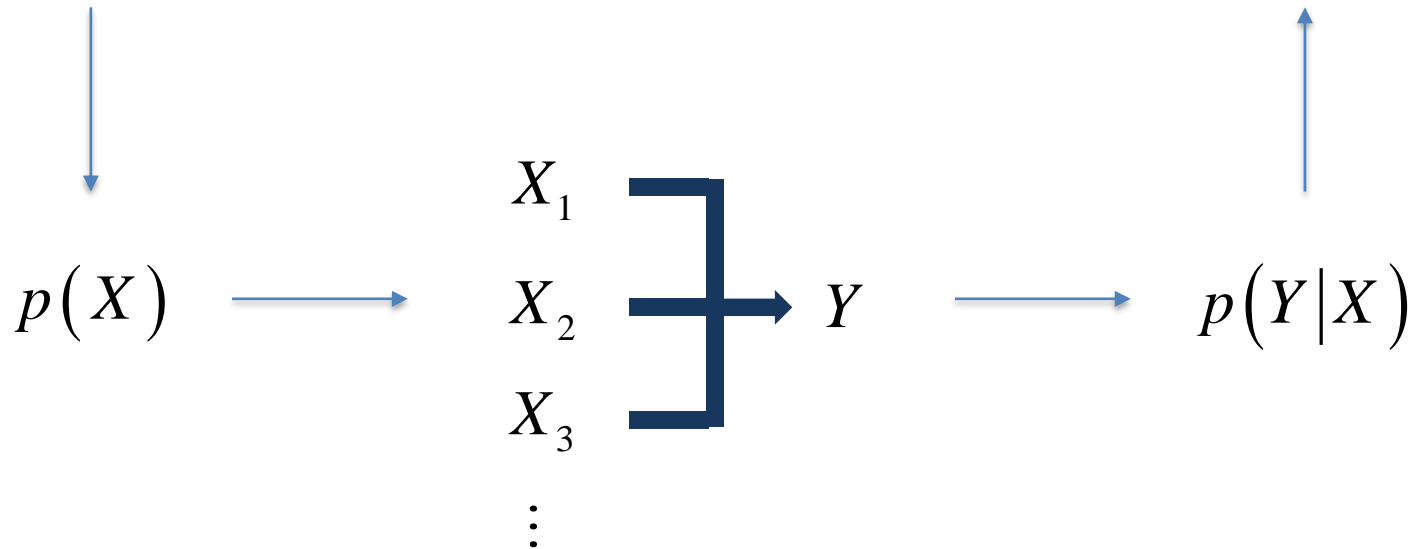
# Intuition



# Setting

$$X \equiv \{X_1, X_2, \dots\}$$

How much synergy in Y about X



# Proposed definition: “synergistic”

Reminder:  $X \equiv \{X_1, X_2, \dots\}$ .

$S_j$  is fully synergistic about  $X$  iff

$$I(S_j : X) > 0,$$

$$I(S_j : X_i) = 0.$$

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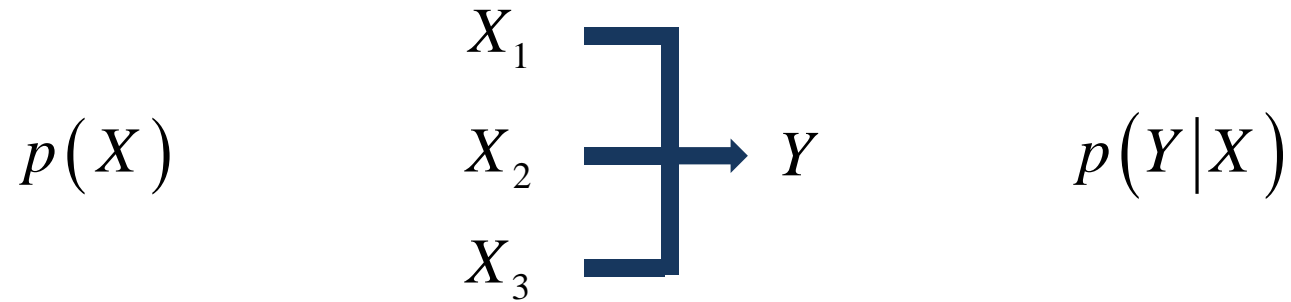
$$I(S_j : X) > 0,$$

$$I(S_j : X_i) = 0.$$

$$S \equiv \{S_j\}_j$$



# First intuition



$$\text{"synergy"} = I(Y : S)^?$$

# Counterexample $S_1, S_2 \in \mathcal{S}$

$$X = \{X_1, X_2\}$$

$$X_i \in \{0, 1, 2\}$$

$$\Pr(X) = 1/9$$

$$S_1 = \begin{cases} (X_1 + X_2) \bmod 3 = 0: & 1 \\ \text{otherwise:} & 0 \end{cases}$$

$$S_2 = \begin{cases} X_1 = (X_2 + 2) \bmod 3: & 1 \\ \text{otherwise:} & 0 \end{cases}$$

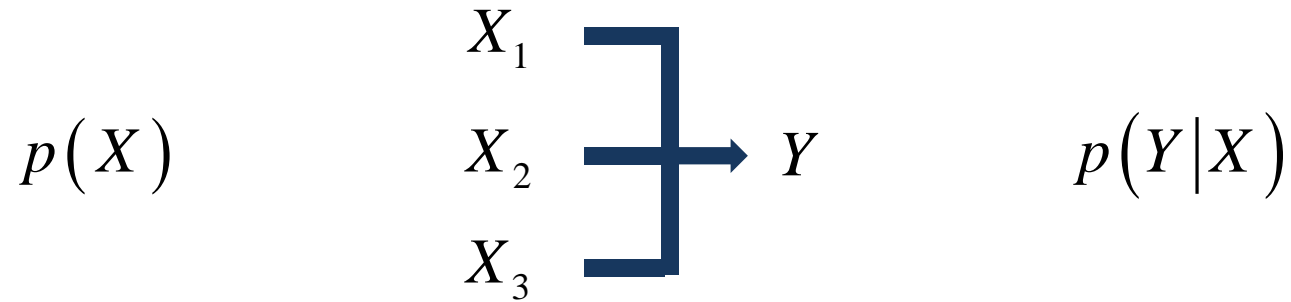
$$I(S_1 : X_1) = 0$$

$$I(S_2 : X_1) = 0$$

$$I(S_1, S_2 : X_1) \approx 1.22$$

$Y=X_1$  would be synergistic?!

# Second intuition



$$\text{"synergy"} \stackrel{?}{=} \sum_j I(Y : S_j)$$

# Counterexample

$$X \equiv \{X_1, X_2, X_3\}$$

$$X_i \in \{0, 1\}$$

$$\Pr(X)$$

$$S_1 = X_1 \oplus X_2$$

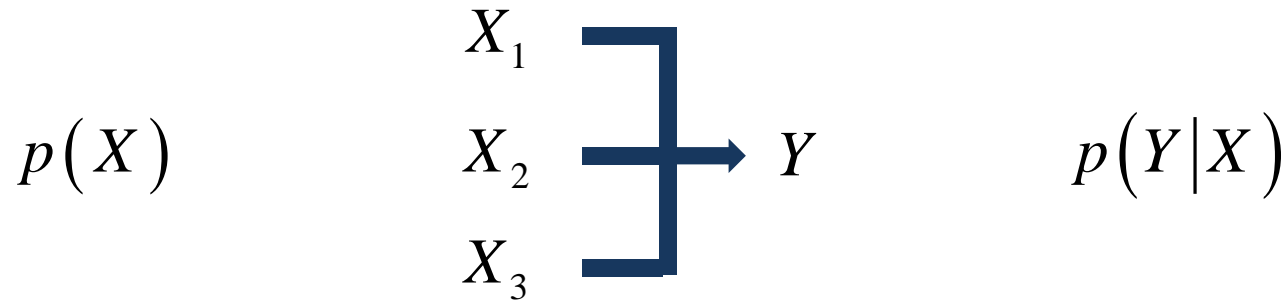
$$S_2 = X_2 \oplus X_3$$

$$S_3 = X_1 \oplus X_3$$

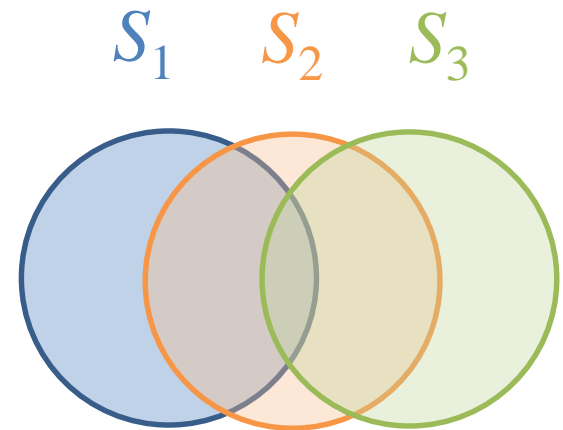
$$S_4 = X_1 \oplus X_2 \oplus X_3$$

Choosing  $Y=\{S_1, S_2\}$  would result in 3 bits of synergy

Why fails?  $\rightarrow S_j$  overlap!

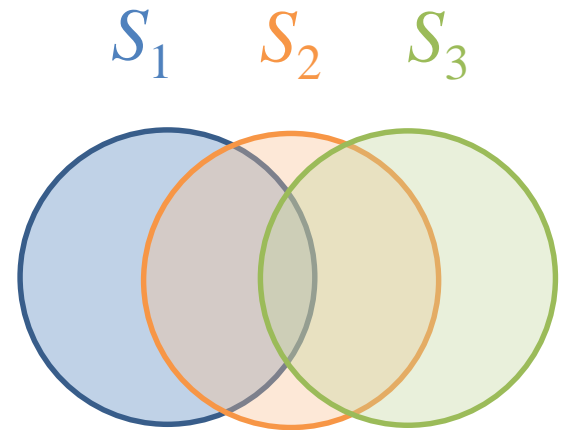


"synergy" =  $\sum_j^? I(Y : S_j)$



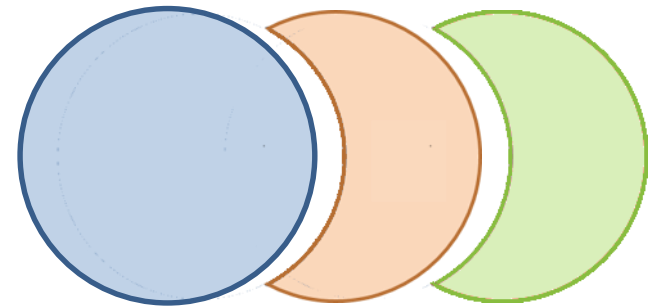
# Final result

$S$



$$S^* \equiv \left( S_i^\perp : S_i^\perp \in D(S_i; S_1, \dots, S_{i-1}) \right)_i,$$

For a given ordering of  $S_i$



$$I_{\text{syn}}(X \rightarrow Y) \equiv \max_{\text{ordering for } S^*} \sum_{S_i \in S^*} I(Y : S_i).$$



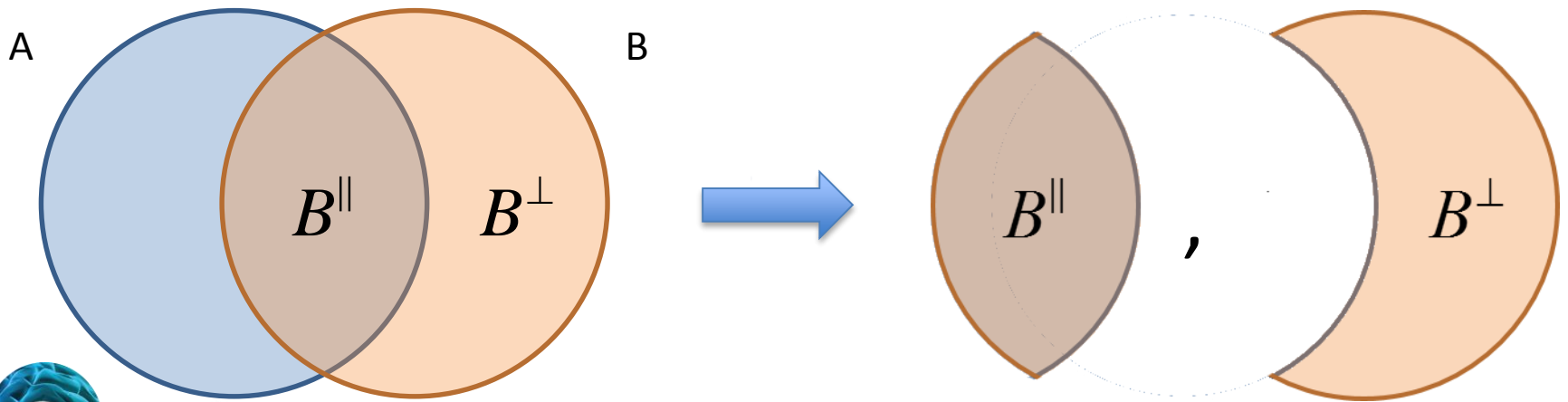
# Orthogonal decomposition

$D : B \mapsto B^\perp, B^\parallel$  such that

$$I(B^\perp, B^\parallel : B) = H(B),$$

$$I(B^\perp : A) = 0,$$

$$I(B^\parallel : A) = I(B : A).$$



# Proved consequences

No 'overcounting'

$$I_{\text{syn}}(X \rightarrow Y) \leq H(S)$$

No 'undercounting'



$$\forall X, \exists Y : I_{\text{syn}}(X \rightarrow Y) = H(S)$$

$I(Y : S_3)$  not counted  $\Rightarrow$  already counted



# Further proved consequences

$$I_{\text{syn}}(X \rightarrow Y) \geq 0.$$

$$I_{\text{syn}}(X \rightarrow Y) \leq I(X : Y).$$

$$I_{\text{syn}}\left(\left(X_i\right)_i \rightarrow \left(Y_i\right)_j\right) = I_{\text{syn}}\left(\left(X_{i'}\right)_{i'} \rightarrow \left(Y_{i'}\right)_{i'}\right).$$

$$I_{\text{syn}}(X_1 \rightarrow Y) = 0.$$

$$I_{\text{syn}}(X \rightarrow X_1) = 0.$$

$$I_{\text{syn}}(X \rightarrow X) = H\left(\sigma_{(S_i)_i}^\perp(X)\right) = H(\sigma(X)) = \max.$$

$$I(X : Y) \leq H(X_1, \dots, X_N) - \max_i H(X_i), \text{ where } Y \in \sigma(X).$$

# Python package: jointpdf

<https://bitbucket.org/rquax/jointpdf>

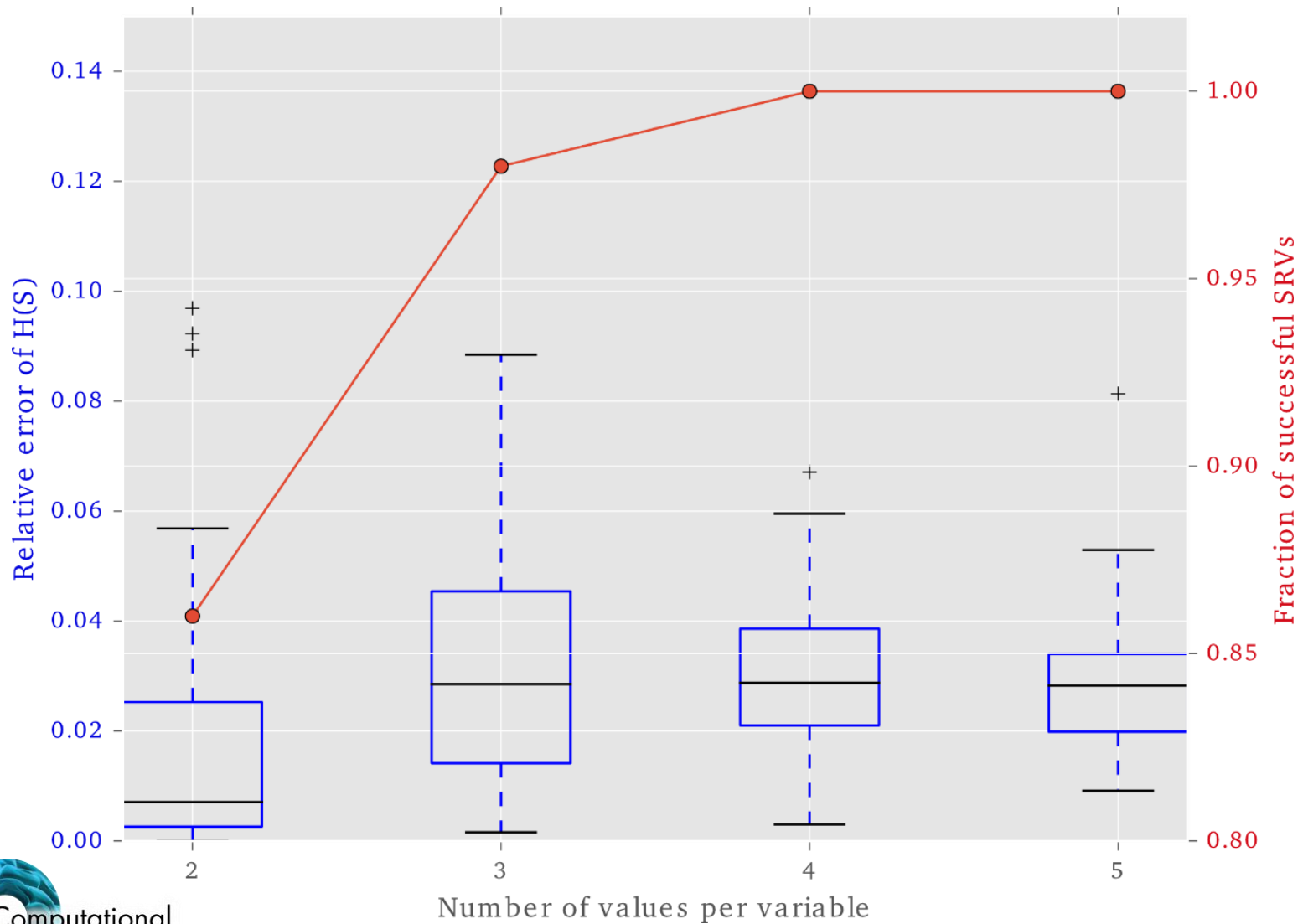
```
from jointpdf import JointProbabilityMatrix

# randomly generate a joint probability mass function p(A,B,C)
# of 3 discrete stochastic variables, each having 4 possible values
p_ABC = JointProbabilityMatrix(3,4)

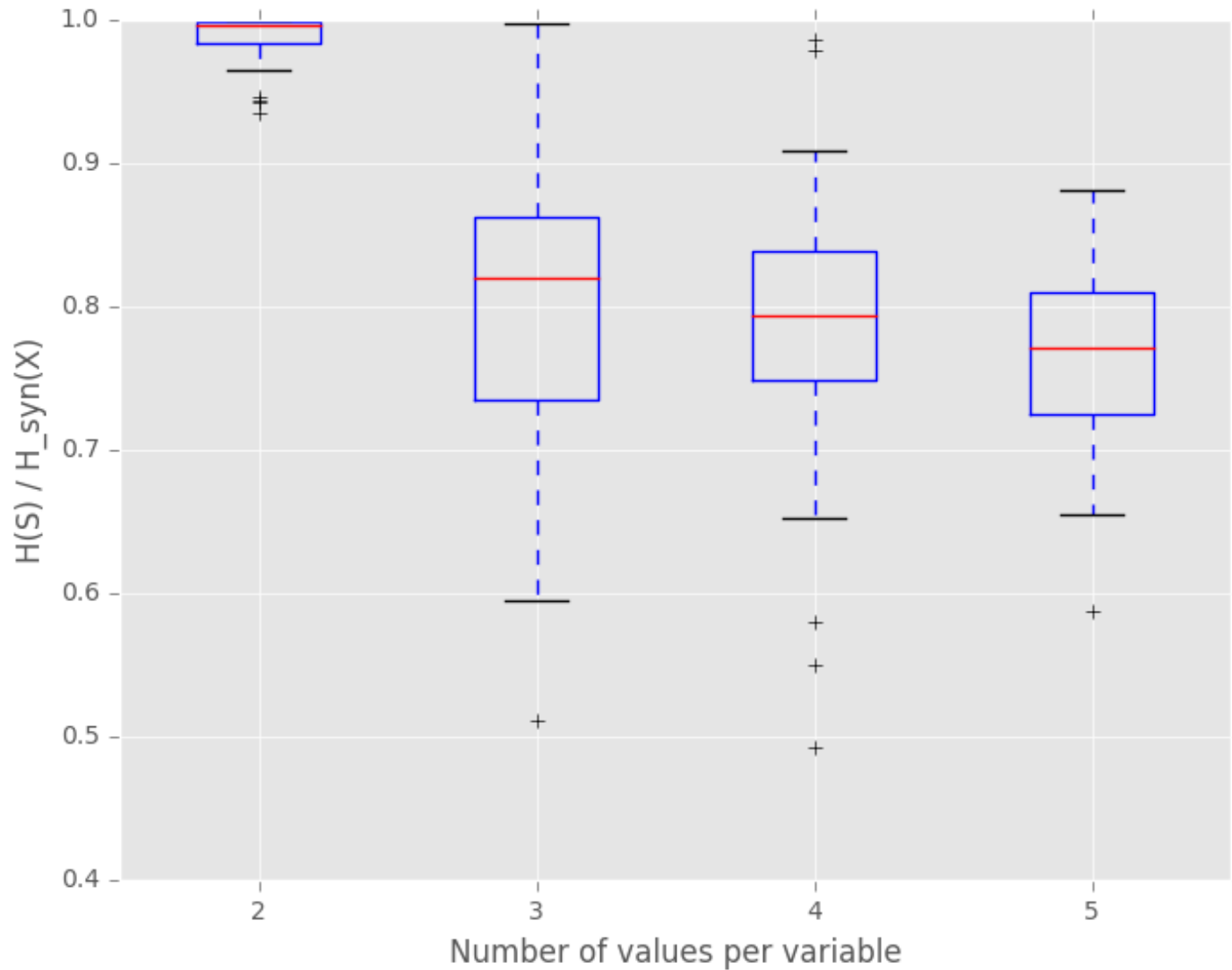
# compute e.g. mutual information between I(A:B)
p_ABC.mutual_information([0], [1])

# compute the information synergy that C contains about A and B (takes a while)
p_ABC.synergistic_information([2], [0,1])
```

# Numerical experiments

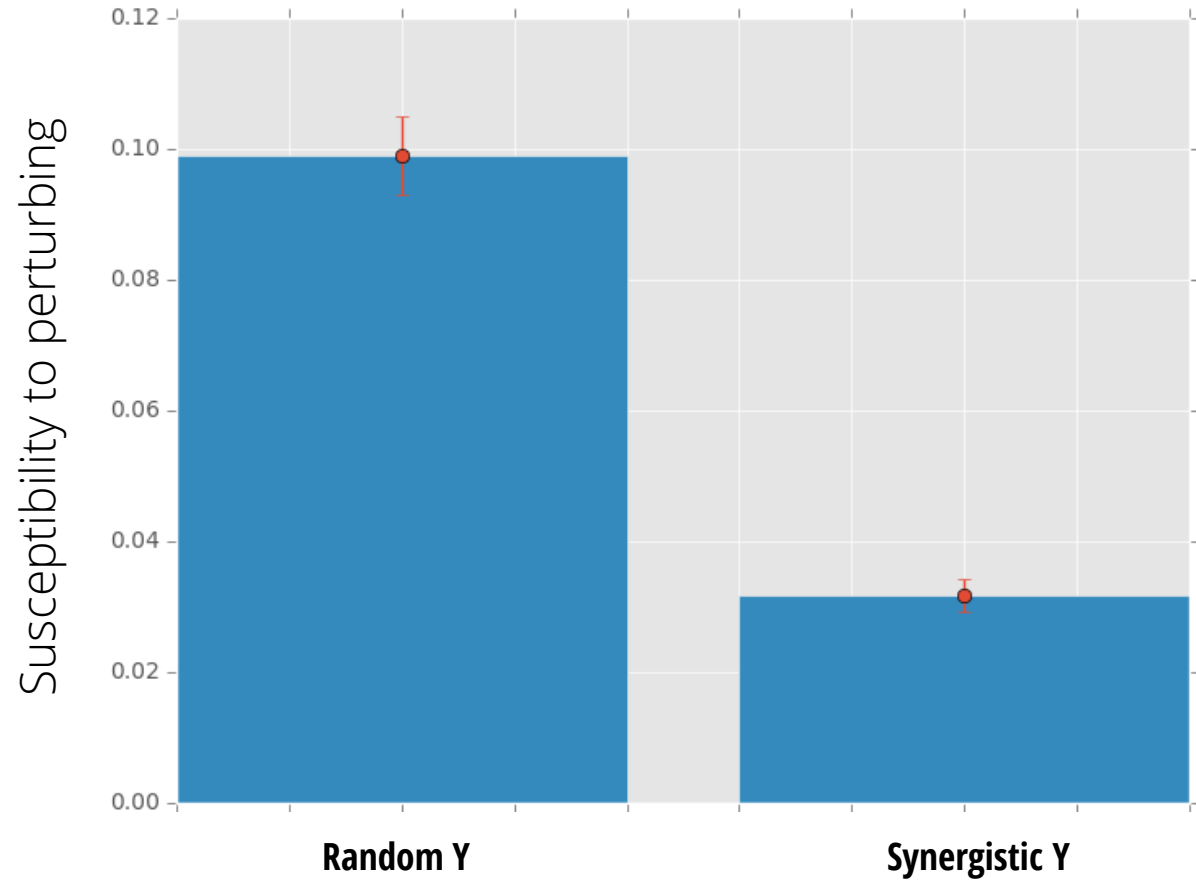
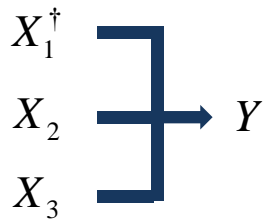


# Numerical experiments



# Operational meaning of synergy

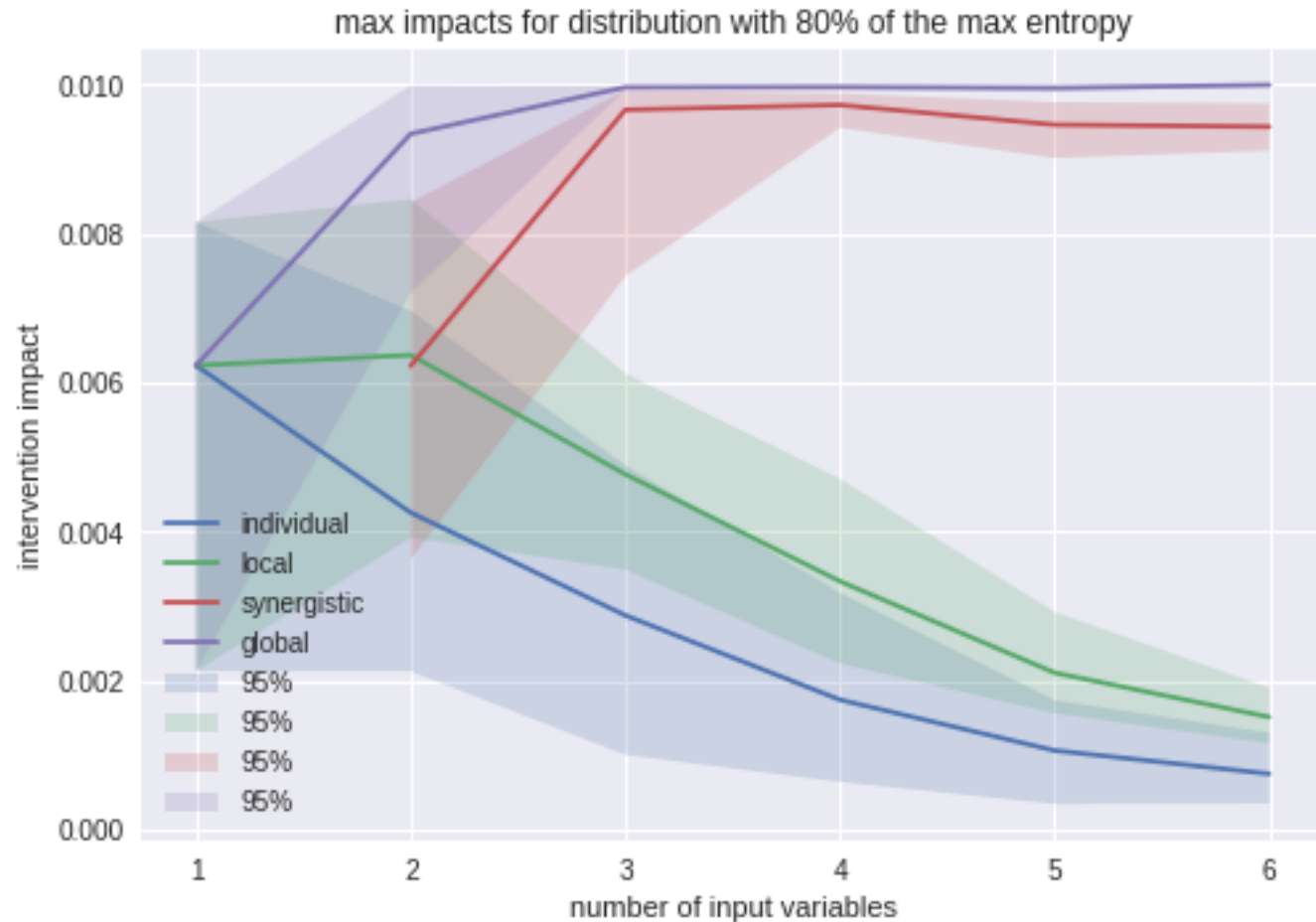
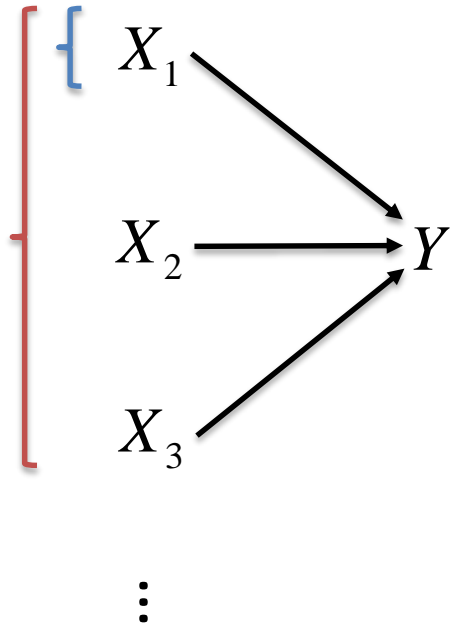
$X_1 \rightarrow X_1^\dagger$



# One-to-one causality?



Derk-Jan Riesthuis



# Information $\rightarrow$ system behavior

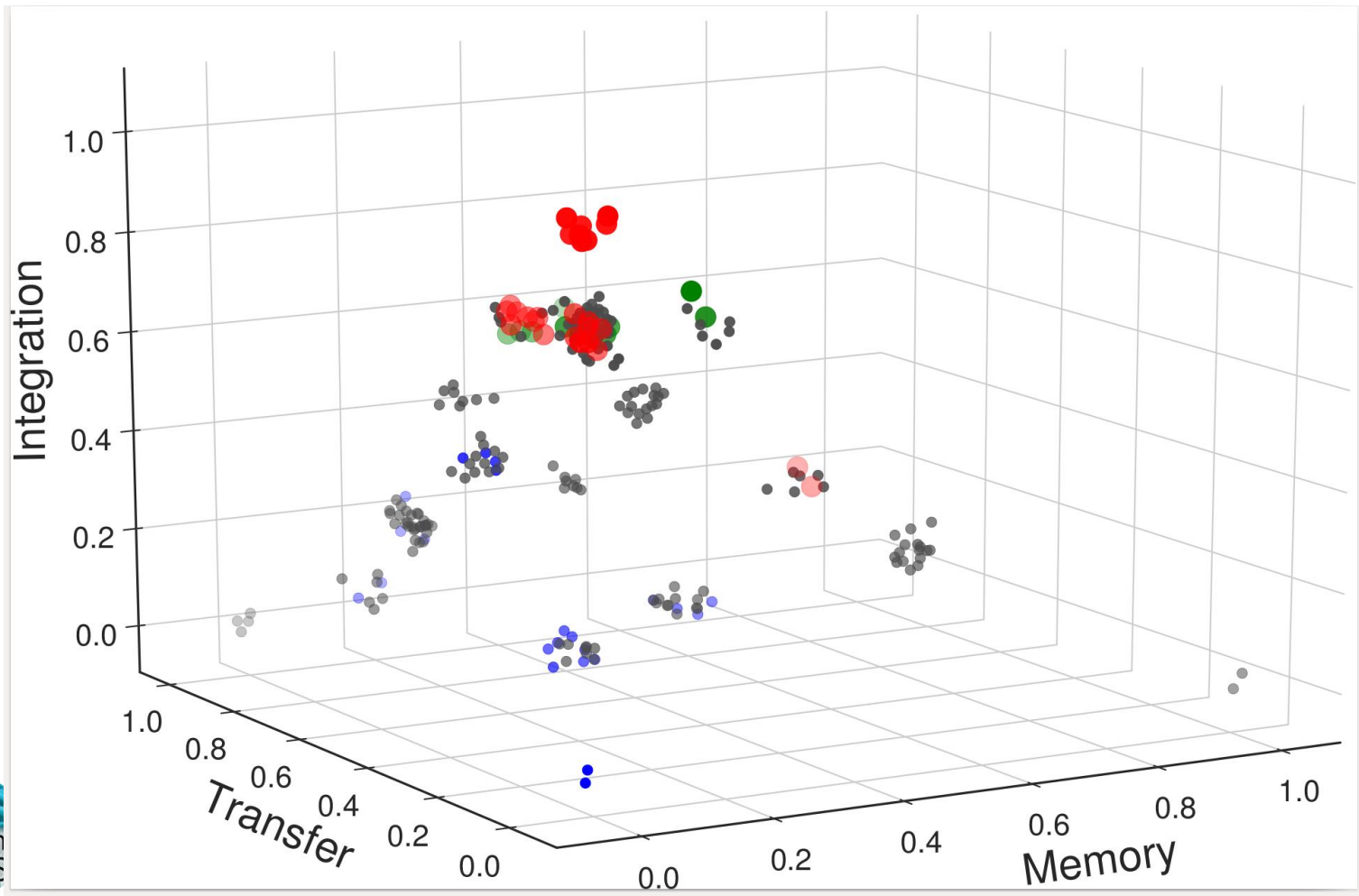
● = static

● = cyclic

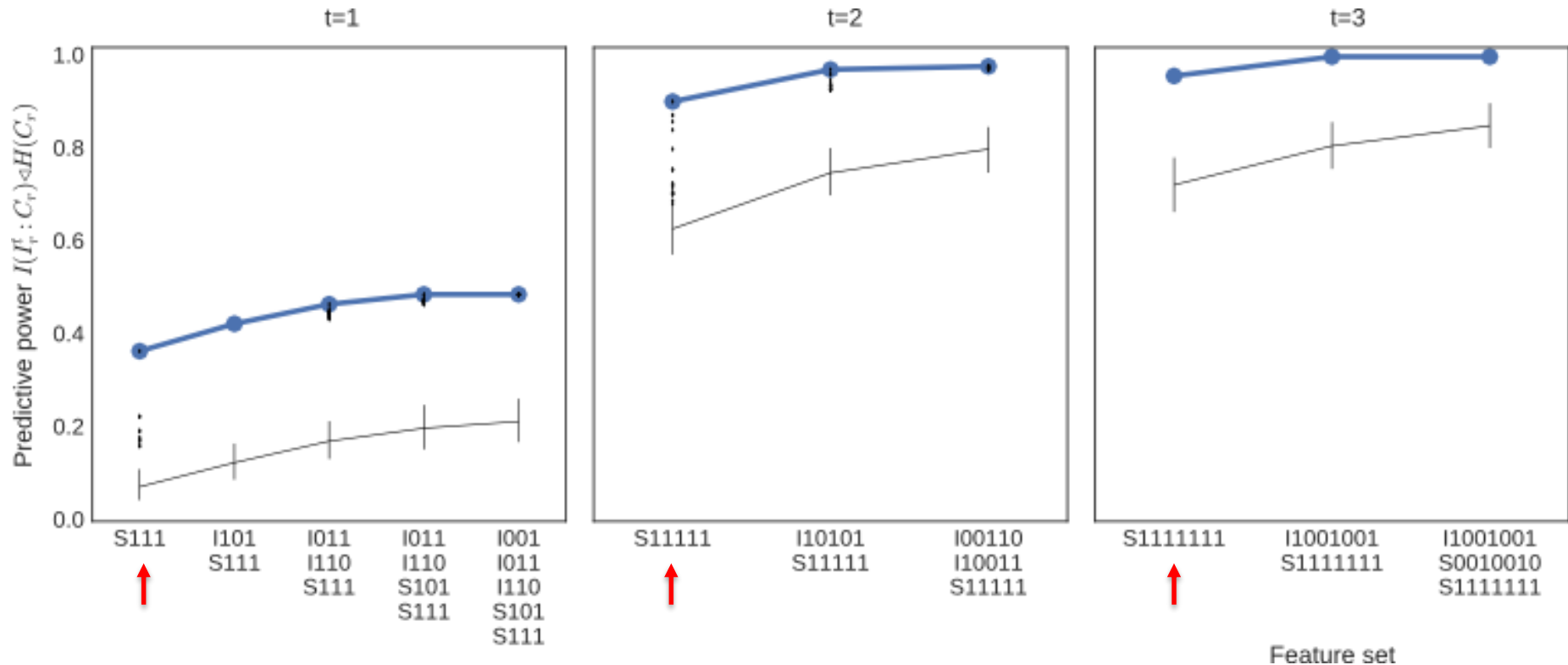
● = chaotic

● = complex

$$I_{\text{syn}} \left( \{X_i\}_i \rightarrow Y \right)$$



# Predictability of Wolfram class



The crucial observation is the “S11..11” feature (“synergy”):

- It is always the single most predictive feature
- The predictability of Wolfram class goes very quickly to 1 (=perfect) with time





# Conclusion

- We use two simple ingredients:
  - Definition of 'synergistic' variable  $S$
  - Correlate  $Y$  with intermediate  $S$  instead of directly  $X$
- $I(X : Y) \neq I_{\text{indiv}} + I_{\text{syn}}$  (=incompatible with PID)
- Limitation: computationally very expensive

Sophocles

topdrim

Software: <https://bitbucket.org/rquax/jointpdf>

# Idea for “individual” information?

