# An operational characterization of mutual information in algorithmic information theory 

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## Two strings $x, y$, and the information therein

$C(x)=$ length of a shortest description of $x$. $C(x \mid y)=$ length of a shortest description of $x$
 given $y$.

Mutual information of $x$ and $y$ is defined by a formula:
$I(x: y)=C(x)+C(y)-C(x, y)$.
Also, $I(x: y)={ }^{+} C(x)-C(x \mid y)$,
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Does $I(x: y)$ have an operational meaning?

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- This was known in the setting of Information Theory (Shannon entropy, etc.) for memoryless and stationary ergodic sources.
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- This was known in the setting of Information Theory (Shannon entropy, etc.) for memoryless and stationary ergodic sources.
- We show it in the framework of Kolmogorov complexity (it has been an open folklore question since the '70s).
- We also have analog results for multiparty secret key agreement protocols.
- We present matching upper/lower bounds for the communication complexity of 2-party secret key agreement protocols, in the public randomness model.


## IT vs. AIT

IT (à la Shannon)

- Data is the realization of a random variable $X$.
- The model: a stochastic process generates the data.
- Amount of information in the data: $H(X)$ (Shannon entropy).

AIT (Kolmogorov complexity)

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## Kolmogorov complexity

Fix $U$ a universal Turing machine.
$p$ is a description of $x$ if $U(p)=x, p$ is a description of $x$ given $y$ if $U(p, y)=x$.
$C(x)=\min \{|p| \mid p$ is a description of $x$.
$C(x \mid y)=\min \{|p| \mid p$ is a description of $x$ given $y$.

## Secret key agreement protocol: warm-up example

- Alice and Bob want to agree on a secret key.
- Problem is that they can only communicate through a public channel.
- Alice knows line $L: y=a_{1} x+a_{0}$;

Bob knows point $P:\left(b_{1}, b_{2}\right)$;

- $L: 2 n$ bits of information (intercept, slope in $\mathbb{F}_{2^{n}}$ ).

- $P: 2 n$ bits of information (the 2 coord. in $\mathbb{F}_{2^{n}}$ ).
- Total information in $(L, P)=3 n$ bits; mutual information of $L$ and $P=n$ bits.


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## SOLUTION:

- Alice sends $a_{1}$ to Bob.
- Bob, knowing that $P \in L$, finds $L$.
- Alice and Bob use $a_{0}$ as a secret key.
- It works! Eve has seen $a_{1}$, but $a_{1}$ and $a_{0}$ are independent.


## Main result (informally stated)

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Theorem (Characterization of the mutual information)
(1) There is a protocol that for every $n$-bit strings $x$ and $y$ allows to compute with high probability a shared secret key of length I(x:y) (up to $-O(\log n))$.
(2) No protocol can produce a longer shared secret key (up to $+O(\log n)$ ).

## Main result (positive part).

## Theorem

There exists a secret key agreement protocol with the following property: if

- Alice knows $x, \epsilon$, and the complexity profile of $(x, y)$,
- Bob knows $y, \epsilon$, and the complexity profile of $(x, y)$, then with probability $1-\epsilon$ they obtain a string $z$ such that,
$|z| \geq I(x: y)-O(\log (n / \epsilon))$
and $C(z \mid$ transcript $) \geq|z|-O(\log (1 / \epsilon))$.


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## Secret key agreement: sketch of the general protocol

- Alice and Bob want to agree on a secret key.
- they can only communicate through a public channel.
- Alice knows : $x$; Bob knows a point $y$;
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## Main result (negative part).

## Theorem

Let $x$ and $y$ be input strings of length $n$ on which the protocol succeeds with error probability $\epsilon$ so that with prob $1-\epsilon$ Alice and Bob have at the end the same $z$, and $C(z \mid t) \geq|z|-\delta(n)$.
Then with probability $\geq 1-O(\epsilon)$ we have $|z| \leq I(x: y)+\delta(n)+O(\log (n / \epsilon))$.

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- technical lemma: $C($ transcript $\mid x)+C($ transcript $\mid y) \leq C($ transcript $)$


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## Result (2): Multi-party secret agreement



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Bob: $\quad x_{2}$
Charlie: $\quad x_{3}$
points $x_{1}, x_{2}, x_{3}$ belong to one line in the affine plane over $\mathbb{F}_{2^{n}}$
maximal common secret key: $n / 2$ bits

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We have $\ell$ parties, given inputs $x_{1}, \ldots, x_{\ell}$.
Each party also knows the complexity profile of $\left(x_{1}, \ldots, x_{\ell}\right)$.

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Under the hood:

- the size of the secret key $\rightarrow$ linear program $\rightarrow$ explicit (but complex) formula [Chan et. al., 2015]
- ... + the same techniques as for $\ell=2$


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Fact: Our protocol for secret key agreement produces a key of length $\approx I(x: y)$ and has communication complexity $\approx \min \{C(x \mid y), C(y \mid x)\}$.

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## Under the hood:

- common information is far less than mutual information; (Gács \& Körner 1970s ; Kolmogorov seminar in 1990s; Muchik \& A.R. 2000s)
- opposition stochastic/nonstochastic objects (Shen 1983; Razenshteyn 2011)


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## Open question

What is the communication complexity for the model with private random bits?

## Previous results: Shannon framework

- Ahlswede and Csiszár [1993] and Maurer [1993]: the optimal size of the common secret key for two parties
- Csiszár and Narayan [2004]: the optimal size of the common secret key for $\ell>2$ parties
- Tyagi [2013]: communication complexity of the protocols
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## formal difference

previous works: random variables \& Shannon's entropy our work: binary strings \& Kolmogorov complexity

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1st substantial difference
previous works: $(\mathrm{X}, \mathrm{Y})$ from random memoryless / stationary ergodic sources our work : no specific structure on X and Y

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## 2nd substantial difference

previous works: protocols work for most admissible pairs $(X, Y)$ our work: protocols work for all admissible pairs $(X, Y)$

## Previous results: Kolmogorov complexity framework

- Finding an operational characterization of mutual information has been a folklore open problem.
- Some earlier approaches:
- Common information of $x$ and $y$ : longest $z$ that can be computed from $x$ and, separately, from $y$ with a few help bits:

$$
C(z \mid x)=O(\log n) \quad C(z \mid y)=O(\log n)
$$

- Is common information equal to mutual information?
- Gács and Körner [1973]: NO!They exhibit $x, y$ with $I(x: y)=\Omega(n)$ and $|z|=o(n)$.
- Muchnik, Romashchenko, Chernov, Vereschagin: several papers with refinements of Gács and Körner.

For example: there are $x, y$ with $C(x), C(y)=^{+} n, I(x: y)=0.99 n,|z|=O(\log n)$.

## One proof

- Upon conditioning, mutual information can increase or decrease.
- There are $x, y, t_{1}, t_{2}$ such that $I\left(x: y \mid t_{1}\right)>I(x: y)$ and $I\left(x: y \mid t_{2}\right)<I(x: y)$
- We show:

If $t(x, y)$ is a function with the rectangle property, then conditioning with $t(x, y)$ decreases mutual information.

- $t$ has the rectangle property, if

$$
t\left(x_{1}, y_{1}\right)=t\left(x_{2}, y_{2}\right)=t \Rightarrow t\left(x_{1}, y_{2}\right)=t .
$$

- This fact is the key point in showing the negative part of our main result.
- The transcript $t(x, y)$ of a protocol on input $(x, y)$ has the rectangle property.
- It has other applications in comm. complexity, maybe elsewhere as well.


## One proof(2)

## Theorem

If $t$ has the rectangle property, then for all $x, y, I(x: y \mid t(x, y)) \leq I(x: y)+O(\log n)$.
Proof:

- Fix $x, y, t=t(x, y)$
- $x^{\prime}$ is a clone of $x$, if $\exists y^{\prime}, t\left(x^{\prime}, y^{\prime}\right)=t$ and $C\left(x^{\prime}\right) \leq C(x)$.
- $y^{\prime}$ is a clone of $y$, if $\exists x^{\prime}, t\left(x^{\prime}, y^{\prime}\right)=t$ and $C\left(y^{\prime}\right) \leq C(y)$.
- Clones $x$ : set of clones of $x$; Clonesy: set of clones of $y$.
- FACT: $\log \mid$ Clones $_{x} \mid \geq C(x \mid t)-O(\log n)$ (and similarly for Clones $y_{y}$.)

Proof: $x$ is described by its ordinal in an enumeration of Clones $_{x}$, which can be done effectively given $t$ and $C(x)$.
So, $C(x \mid t) \leq \log \mid$ Clones $_{x} \mid+O(\log n)$.

## One proof(3)

## Theorem

If $t$ has the rectangle property, then for all $x, y, I(x: y \mid t(x, y)) \leq I(x: y)+O(\log n)$.

Proof (continuation):

- Take a pair $\left(x^{\prime}, y^{\prime}\right) \in$ Clones $_{x} \times$ Clones $_{y}$ with maximal $C\left(x^{\prime}, y^{\prime} \mid t\right)$.
- $C\left(x^{\prime}, y^{\prime} \mid t\right) \geq^{+} \log \mid$ Clones $_{x} \times$ Clones $_{y} \mid={ }^{+} C(x \mid t)+C(y \mid t)$.
- $C\left(x^{\prime}, y^{\prime}\right)=C\left(x^{\prime}, y^{\prime}, t\right) \quad$ (because $t=t\left(x^{\prime}, y^{\prime}\right)$, using the rectangle property) $={ }^{+} C(t)+C\left(x^{\prime}, y^{\prime} \mid t\right) \quad$ (chain rule)
$\geq^{+} C(t)+C(x \mid t)+C(y \mid t)$.
- On the other hand,

$$
C\left(x^{\prime}, y^{\prime}\right) \leq C\left(x^{\prime}\right)+C\left(y^{\prime}\right) \leq C(x)+C(y) \text { (by def. of clones) }
$$

- Combining the last two inequalities, $C(t)+C(x \mid t)+C(y \mid t) \leq^{+} C(x)+C(y)$.
- Now subtract $C(x, y, t)$ in the LHS, and (the smaller) $C(x, y)$ in the RHS.
- QED


## Take home message

- Operational characterization of the mutual information of strings $x$ and $y$ :
$I(x: y)$ is equal (up to logarithmic precision) to the length of a longest secret key that two parties, one having $x$ and the other having $y$, can establish via an interactive protocol on an open channel.

The protocol is probabilistic and the parties also need to know how their strings are correlated (i.e., they know the complexity profile of $x$ and $y$ ).

- The protocol has communication complexity $\min (C(x \mid y), C(y \mid x))$.
- The communication is optimal for finding a secret key of maximal length, in the model with public randomness.
- We also determine the maximum length of a shared secret key in the multi-party setting.


## Thank you

Full version:

- A. Romashchenko and M. Zimand, An operational characterization of mutual information in algorithmic information theory, available at ECCC https://eccc.weizmann.ac.il/report/2018/043

