Logical Induction

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This talk is based on our paper,

http://arXiv.org/abs/1609.03543/

which will be updated more frequently at

https://intelligence.org/files/LogicalInduction.pdf

These slides will be available at:

https://intelligence.org/seminar-f2016/
and possibly in a more updated form at:
http:/acritch.com/research/

Overview

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Formalizing logical induction

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Definitions

- $\mathcal{L} :=$ a **language** of propositional logic, including connectives \neg , \land , \lor , \rightarrow , \leftrightarrow , for constructing proofs using modus ponens.
- $\mathcal{S} :=$ all **sentences** expressible in \mathcal{L} .
- Γ := a set of axioms in S for encoding and proving statements about variables and computer programs (e.g. First Order Logic + Peano Arithmetic).
- a **belief state** := a map $\mathbb{P} : S \to [0, 1]$ that is constant outside some finite subset of S.
- a reasoning process P
 := a computable sequence of belief states {P_n : L → [0, 1]}.

We can now state some properties that we think a "good reasoning process" should satisfy.

Basic properties

- A "good" reasoning process $\overline{\mathbb{P}}$ should satisfy:
 - (computability) There should be a Turing machine which computes $\mathbb{P}_n(\phi)$ for any input (n, ϕ) .
 - (convergence) The limit P_∞(φ) := lim_{n→∞} P_n(φ) should exist for all sentences φ.
 - (coherent limit) P_∞ should be a coherent probability distribution, i.e. obey laws like
 P_∞(A ∧ B) + P_∞(A ∨ B) = P_∞(A) + P_∞(B)
 - (non-dogmatism) If Γ ⊭ φ then P_∞(φ) < 1, and if Γ ⊭ ¬φ then P_∞(φ) > 0.

Progress

Our paper (http://arXiv.org/abs/1609.03543/), shows that these properties are:

Related: A single property, the **Garrabrant Induction Criterion** (GIC), implies them all.

Feasible: We have a logical induction algorithm, "LIA2016", that satisfies the GIC.

Extensible: Many further desirable properties follow from **GIC**, and are hence satisfied by **LIA2016**.

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Conservatism

(uniform non-dogmatism) For any computably enumerable sequence of sentences {φ_n}_{n∈N} such that Γ ∪ {φ_n}_{n∈N} is consistent, there is a constant ε > 0 such that for all n,

$$\mathbb{P}_{\infty}(\phi_n) \geq \varepsilon.$$

(Occam bounds) There exists a fixed positive constant C such that for any sentence φ with Kolmogorov complexity κ(φ) in a prefix-free encoding, if Γ ⊬ ¬φ, then

$$\mathbb{P}_{\infty}(\phi) \geq C 2^{-\kappa(\phi)},$$

and if $\Gamma \nvDash \phi$, then

$$\mathbb{P}_{\infty}(\phi) \leq 1 - C 2^{-\kappa(\phi)}.$$

(definition: efficiently computable)

We say that a sequence of statements (or other objects) $\overline{\phi}$ is **efficiently computable (e.c.)** if there exists a Turing machine M such that M(n) generates the output ϕ_n in time polynomial in n.

An e.c. sequence ϕ_n can be thought of as a sequence of T/F questions that is relatively easy to generate, but which can be arbitrarily difficult to answer deductively as *n* grows. In other words, think:

e.c. statements

 \leftrightarrow

easy to state, hard to verify

Henceforth, $\overline{\phi}$ will always denote an e.c. sequence of sentences.

(definition: efficiently computable)

Example (statements that are hard to verify). Say f is any computable function. Fix an encoding \underline{f} of f. By the parametric diagonal lemma [Boolos, 1993; p.53], there is a sentence G(-) with one free variable such that for all n, Γ proves

 $G(\underline{n}) \leftrightarrow$ "There is no proof of $\underline{G(\underline{n})}$ in $\leq \underline{f(\underline{n})}$ characters."

Then the sequence $\phi_n := G(\underline{n})$ is log-time generable: writing down ϕ_n only requires substituting the string \underline{n} into G(-), which takes $\mathcal{O}(\log(n))$ time. But if Γ is consistent, the length of the shortest proof of ϕ_n is at least f(n). Nonetheless, we have...

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Provability induction

• (provability induction) For any e.c. sequence $\overline{\phi}$ of provable statements ϕ_n ,

 $\lim_{n\to\infty}\mathbb{P}_n(\phi_n)=1.$

In particular, $\overline{\mathbb{P}}$ can be seen to "outpace deduction" by a factor of f for any computable function f.

An analogy: Ramanujan vs Hardy. Imagine the ϕ_n are output by a heuristic algorithm that generates mathematical facts without proofs, similar in style to S. Ramanujan. Then $\overline{\mathbb{P}}_n$ resembles G.H. Hardy: he can only verify those results very slowly using the proof system Γ , but after enough examples, he begins to trust Ramanujan as soon as he speaks, even if the proofs of Ramanujan's later conjectures are impossibly long.

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Learning pseudorandom frequencies

In the paper, we define a notion of *pseudorandom* with respect to a particular runtime class $\mathcal{O}(r(n))$ depending on the runtime of $\overline{\mathbb{P}}$. Black-boxing those for now, we have:

(Learning pseudorandom frequencies) For any e.c. sequence of decidable sentences φ
 that is pseudorandom with frequency p over the class of O(r(n))-time divergent weightings,

$$\lim_{n\to\infty}\mathbb{P}_n(\phi_n)=p.$$

• (Learning pseudorandom trends) A stronger version of the above, where the frequency can vary over time.

Learning pseudorandom frequencies

Note that learning pseudorandom frequencies

- is not that hard to satisfy on its own, but
- is trickier to get along with coherence (i.e., P_∞ being a probability distribution).

Learning provable relationships

 (Learning exclusive/exhaustive relationships) Let
 φ¹,..., φ^k be k e.c. sequences of sentences such that for each
 n, Γ proves that φ¹_n,..., φ^k_n are exclusive and exhaustive (i.e.
 exactly one of them is true). Then

$$\lim_{n\to\infty} \left(\mathbb{P}_n(\phi_n^1) + \cdots + \mathbb{P}_n(\phi_n^k) \right) = 1$$

• (Learning affine relationships) A stronger version of the above, holding for every coherence relationship expressible as an affine combination of probabilities.

(definition: timely manner)

Given any sequences \overline{x} and \overline{y} , we write

$$\begin{array}{ll} x_n \simeq_n y_n \quad \text{for} \quad \big(\lim_{n \to \infty} x_n - y_n = 0\big), \\ x_n \gtrsim_n y_n \quad \text{for} \quad \big(\liminf_{n \to \infty} x_n - y_n \ge 0\big), \text{ and} \\ x_n \lesssim_n y_n \quad \text{for} \quad \big(\limsup_{n \to \infty} x_n - y_n \le 0\big). \end{array}$$

Given e.c. sequences of statements $\overline{\phi}$ and probabilities \overline{p} , we say that $\overline{\mathbb{P}}$ assigns \overline{p} to $\overline{\phi}$ in a **timely manner** if

$$\mathbb{P}_n(\phi_n) \simeq_n p_n$$

Self-reflective properties

(introspection) For any efficiently computable sequence of statements φ_n, any interval (a, b), any e.c. sequence of positive rationals δ_n → 0, there exists a sequence ε_n → 0 such that for all n:

$$\mathbb{P}_n(\phi_n) \in (\mathbf{a} + \delta_n, \mathbf{b} - \delta_n) \implies \mathbb{P}_n(\lceil \mathbb{P}_n(\phi_n) \in (\mathbf{a}, \mathbf{b})\rceil) > 1 - \varepsilon_n$$

$$\mathbb{P}_n(\phi_n) \notin (\mathbf{a} - \delta_n, \mathbf{b} + \delta_n) \implies \mathbb{P}_n(\lceil \mathbb{P}_n(\phi_n) \notin (\mathbf{a}, \mathbf{b})\rceil) < \varepsilon_n$$

 (paradox resistance) Fix a rational p ∈ (0, 1), and use Gödels diagonal lemma to define a sequence of "Liar sentences" L_n satisfying

$$\Gamma \vdash L_n \leftrightarrow \lceil \mathbb{P}_n(L_n) \leq p \rceil.$$

Then

$$\overline{\mathbb{P}}_n(L_n)\simeq_n p.$$

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$$\mathbb{P}_n(\phi_n) \in (\mathbf{a} + \delta_n, \mathbf{b} - \delta_n) \implies \mathbb{P}_n\big(\left[\mathbb{P}_n(\phi_n) \in (\mathbf{a}, \mathbf{b})^{\mathsf{T}} \right] > 1 - \varepsilon_n \\ \mathbb{P}_n(\phi_n) \notin (\mathbf{a} - \delta_n, \mathbf{b} + \delta_n) \implies \mathbb{P}_n\big(\left[\mathbb{P}_n(\phi_n) \notin (\mathbf{a}, \mathbf{b})^{\mathsf{T}} \right] < \varepsilon_n$$

 (paradox resistance) Fix a rational p ∈ (0,1), and use Gödels diagonal lemma to define a sequence of "Liar sentences" L_n satisfying

$$\Gamma \vdash L_n \leftrightarrow {}^{r}\mathbb{P}_n(L_n) \leq p^{r}.$$

Then

$$\overline{\mathbb{P}}_n(L_n)\simeq_n p.$$

Self-reflective properties

(belief in consistency) Let con(n) be the sentence 'There is no proof of contradiction (⊥) from Γ using n or fewer symbols[¬]. Then

 $\lim_{n\to\infty}\overline{\mathbb{P}}_n(\operatorname{con}(n))=1.$

• (belief in future consistency) In fact, for any encoding \underline{f} of a computable function $f : \mathbb{N} \to \mathbb{N}$,

 $\lim_{n\to\infty}\overline{\mathbb{P}}_n(\operatorname{con}(\underline{f}(n)))=1.$

For example, f(n) could be n^{n^n} , or even Ack(n, n).

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Self-reflective properties

• (Trust in future beliefs) For any computable function f(n) > n and efficiently computable sentences ϕ_n , we have a result roughly interpretable as saying that a GI's current beliefs about the sequence, conditioned on its future beliefs, agree with its future beliefs:

$$\mathbb{P}(\phi_n \mid "\underline{\mathbb{P}}_{\underline{f(n)}}(\underline{\phi_n}) \geq \underline{p_n}") \gtrsim_n p_n.$$

The precise statement (see paper for definitions) looks like this:

$$\mathbb{E}_n([\underline{\phi_n}] \cdot \underline{\mathsf{Ind}}_{\delta_n}("\underline{\mathbb{P}}_{\underline{f}(\underline{n})}(\underline{\phi_n}) \geq \underline{p_n}")) \gtrsim_n p_n \cdot \mathbb{E}_n("\underline{\mathbb{P}}_{\underline{f}(\underline{n})}(\underline{\phi_n})").$$

Other properties

- Well-behaved conditional credences, the analog of conditional probabilities;
- Well-behaved *logically uncertain variables*, the analogues of classical random variables;
- Well-behaved expected value operators for logically uncertain variables;
- Relationship to universal semi-measures;
- · · · (check out the paper)

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The Garrabrant induction criterion

A market $\overline{\mathbb{P}}$ is said to satisfy the **Garrabrant induction criterion** relative to a *deductive process* \overline{D} if there is no efficiently computable *trader* \overline{T} that (*plausibly*) *exploits* $\overline{\mathbb{P}}$ relative to \overline{D} . A market $\overline{\mathbb{P}}$ that meets this criterion is called a **Garrabrant inductor**.

A **deductive process** \overline{D} is a computable nested sequence $D_1 \subseteq D_2 \subseteq D_3 \ldots$ of finite sets of sentences $D_n \subset S$, interpreted as theorems that have been proven by day n. We write D_{∞} for the union $\bigcup_n D_n$.

A trader \overline{T} is a sequence of things called *n*-strategies T_n , each of which is a formula for buying and selling a linear combination of "shares" of sentences $T_n(\mathbb{P}_{\leq n})$ in response to the history of market prices $\mathbb{P}_{\leq n}$ on day *n*.

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A trader's (cash and stock) holdings on day *n* from trading against $\overline{\mathbb{P}}$ is the sum $H_n := \sum_{i \leq n} T_n(\mathbb{P}_{\leq n})$.

A trader \overline{T} (plausibly) exploits a market $\overline{\mathbb{P}}$ if, as $n \to \infty$, the bounds on the value of its holdings H_n determinable from D_n via boolean logic only are bounded below but not bounded above.

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Example. Say $\phi = "1 + 1 = 2"$ and $\chi = "2 + 2 = 4"$, and suppose you're a trader whose your holdings on day 5 are

$-\mathbf{1} + \phi + \chi$

representing -\$1 of cash, one share of ϕ and one share of χ .

- If $D_5 = \emptyset$, the current bounds on your worth are [-1, 1].
- If $D_5 = \{\phi\}$, your bounds are [0, 1].
- If $D_5 = \{\phi \land \chi\}$, your bounds are [1, 1] (the \land is respected)
- If D₅ = {∀x : φ}, your bounds are only [-1,1] (the quantifier ∀ is not respected)

The Garrabrant induction criterion

Time permitting, use whiteboard to elaborate and/or field questions.



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The basic ideas behind **LIA2016** are these:

- We fix a (redundant) computable enumeration of all e.c. traders, and define two functions:
- TradingFirm watches a market P_{≤n} and assembles performance-budgeted versions of those traders together, yielding a non-e.c. "supertrader" T who exploits P iff P is exploitable.
- MarketMaker looks at any trading strategy T_n and sets prices so that strategy can't make more than 2⁻ⁿ from trading with them (no matter how stocks are valued).
- LIA pits MarketMaker and TradingFirm against each other in a recursion, which builds a market ₱ not exploitable by the output of TradingFirm applied to it, and hence not by and e.c. trader.

Given the deductive process \overline{D} , the shape of the recursion looks like this: LIA_{≤ 0} := (), and

$$\begin{split} \texttt{LIA}_n := \texttt{MarketMaker}_n(\texttt{TradingFirm}_n^D(\texttt{LIA}_{\leq n-1}),\texttt{LIA}_{\leq n-1}), \\ \text{After enough lemmas and definitions, the main existence result} \\ \texttt{looks like this:} \end{split}$$

Theorem $(\overline{\text{LIA}} \text{ is a Logical Inductor})$

The sequence of belief states $\overline{\text{LIA}}$ satisfies the Garrabrant induction criterion relative to \overline{D} , i.e., $\overline{\text{LIA}}$ is not exploitable by any e.c. trader relative to the deductive process \overline{D} .

Proof.

If any e.c. trader exploits $\overline{\text{LIA}}$ (relative to \overline{D}), then so does the trader $\overline{F} := (\text{TradingFirm}_n^{\overline{D}}(\text{LIA}_{\leq n-1}))_{n \in \mathbb{N}^+}$. But \overline{F} does not exploit $\overline{\text{LIA}}$. Therefore no e.c. trader exploits $\overline{\text{LIA}}$.

Time permitting, use whiteboard to elaborate and/or field questions.



IA2016

LIA2016

The proofs of all our nice properties involve cooking up some e.c. trader that would exploit you otherwise. E.g.:

Proof sketch of Convergence.

Suppose for a contradiction that the limit

$$\mathbb{P}_{\infty}(\phi) := \lim_{n \to \infty} \mathbb{P}(\phi)$$

does not exist. Then for some rationals $p \in [0,1]$ and $\varepsilon > 0$, we have $\mathbb{P}_n(\phi) and <math>\mathbb{P}_n(\phi) > p + \varepsilon$ infinitely often, so a trader can make ∞ buy buying shares for less than $p - \varepsilon$, waiting for a chance to sell then for $p + \varepsilon$, and repeating (details in paper).

LIA2016

Proof sketch of Non-dogmatism.

Suppose for a contradiction that $\Gamma \nvDash \neg \phi$, but $\mathbb{P}_{\infty}(\phi) = 0$. (The other case is similar.) A trader can buy one share of ϕ at or below every price point 2^{-k} , never spending more than \$1, but accruing an even growing number of ϕ -shares $k \cdot \phi$. Since we never have $D_n \vdash \phi$, those shares are plausibly worth k, which $\rightarrow \infty$ as $n \rightarrow \infty$, contradicting the *GIC*. Hence $\mathbb{P}_{\infty}(\phi)$ must be bounded away from zero.

See the paper for more rigorous details, and many more properties/proofs:

http://arXiv.org/abs/1609.03543/
https://intelligence.org/files/LogicalInduction.pdf

(The latter is being updated more frequently.)

Conclusions (PowerPoint)

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Logical Induction Conclusions (PowerPoint) Andrew Critch (MIRI)

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