HOL Theorem Proving and Formal Probability (2)

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Outline of Contents

- Preliminary (set theory, topology, reals and extended reals, etc.)
- Measure Theory
- Borel Measure Space
- Lebesgue Integration Theory
- Probability Theory
- ☐ Simple Stochastic Processes
 - Random sequences (IID and stationary)
 - Martingale
 - (Markov chain)



History of HOL4-Probability

- ☐ Joe Hurd, Formal verification of probabilistic algorithms, University of Cambridge, UCAM-CL-TR-566, 2003.
- Osman Hasan, Formal Probabilistic Analysis using Theorem Proving, Concordia University (Hardware Verification Group), 2008.
- □ Aaron R. Coble, Anonymity, information, and machine-assisted proof, University of Cambridge, UCAM-CL-TR-785, 2010.
- ☐ Tarek Mhamdi, Information-Theoretic Analysis Using Theorem Proving, Concordia University (Hardware Verification Group), 2012.
- Chun Tian, Carathéodory extension theorem for semirings, Construction of the one-dimensional Borel measure, Product measure, The Law of Large Numbers, etc. 2018-2022 (no publications)



Reference Books

- A. N. Kolmogorov, Foundations of the Theory of Probability (Grundbegriffe der Wahrscheinlichkeitsrechnung). Chelsea Publishing Company, 1950. (orig. 1933)
- □ R. L. Schilling, Measures, Integrals and Martingales, 2nd ed., Cambridge University Press, 2017.
- □ K. L. Chung, A Course in Probability Theory, 3rd ed., Academic Press, 2001.
- J. S. Rosenthal, A First Look at Rigorous Probability Theory, 2nd ed., World Scientific Publishing Company, 2006.
- P. Billingsley, Probability and Measure, 3rd ed., John Wiley & Sons, 1995.
- □ A. N. Shiryaev, Probability-1, 3rd ed., Springer-Verlag, 2016. (orig. 2007)
- A. N. Shiryaev, Probability-2, 3rd ed., Springer-Verlag, 2019.



Family of Sets: Algebra (aka Field)

```
(X, \mathcal{A}) is an algebra (or field) if:
A \in \mathcal{A} \Rightarrow A \subseteq X
\sqcap \emptyset \in \mathcal{A}
    A \in \mathcal{A} \Rightarrow X \setminus A \in \mathcal{A} \text{ (or } A \in \mathcal{A})
A, B \in \mathcal{A} \Rightarrow A \cup B \in \mathcal{A}
 \vdash subset_class sp sts \iff \forall x. \ x \in sts \Rightarrow x \subseteq sp
 \vdash algebra a \iff
      subset\_class (space a) (subsets a) \land
     \emptyset \in \text{subsets } a \land
     (\forall s. \ s \in \text{subsets} \ a \Rightarrow \text{space} \ a \ \text{DIFF} \ s \in \text{subsets} \ a) \ \land
     \forall s \ t.
         s \in \text{subsets } a \land t \in \text{subsets } a \Rightarrow
         s \cup t \in \mathtt{subsets} \ a
                                                                                                     [algebra_def]
X \in \mathcal{A} \text{ (as } X = \emptyset \text{ or } X \setminus \emptyset)
    A, B \in \mathcal{A} \Rightarrow A \cap B \in \mathcal{A} \text{ (as } A \cap B = A \cup B \in \mathcal{A})
    A, B \in \mathcal{A} \Rightarrow A \setminus B \in \mathcal{A} \text{ (as } A \setminus B = A \cap B \in \mathcal{A})
```



System of Sets: σ -Algebra

 (X,\mathcal{A}) is an σ -algebra if (X,\mathcal{A}) is algebra and, additionally:

$$\forall A_1, A_2, \dots \in \mathcal{A} \Rightarrow \bigcup_{i \in \mathbb{N}} A_i \in \mathcal{A}$$

```
\vdash \operatorname{sigma\_algebra} \ a \iff \\ \operatorname{algebra} \ a \ \land \\ \forall \ c. \ \operatorname{countable} \ c \ \land \ c \subseteq \operatorname{subsets} \ a \Rightarrow \\ \bigcup \ c \in \operatorname{subsets} \ a \implies \\ [\operatorname{sigma\_algebra\_def}] \vdash \operatorname{sigma\_algebra} \ (sp, sts) \iff \\ \operatorname{algebra} \ (sp, sts) \ \land \\ \forall \ A. \ \operatorname{IMAGE} \ A \ \mathcal{U}(:\operatorname{num}) \subseteq sts \Rightarrow \\ \bigcup \ \{A \ i \ | \ i \in \mathcal{U}(:\operatorname{num})\} \in sts \qquad [\operatorname{sigma\_algebra\_alt}]
```

Alternatively, $\forall A_1, A_2, ... \in \mathcal{A} \Rightarrow \bigcap_{i \in \mathbb{N}} A_i \in \mathcal{A}$



Trivial $(\sigma$ -)algebras

Some trivial algebras (and also σ -algebras):

- Smallest $(\sigma$ -)algebra generated from any space X: $(X, \{\emptyset, X\})$
- \square Single-set $(\sigma$ -)algebra: $S \subseteq X \Rightarrow (X, \{\emptyset, S, X \setminus S, X\})$
- $lue{\Box}$ Biggest $(\sigma$ -)algebra generated from any space X: $(X, \mathcal{P}(X))$

In particular, finite algebra (finite space or finite subsets) is always σ -algebra:



Extended Real Numbers

extreal = NegInf | PosInf | Normal real

```
[extreal add def]
\vdash Normal x + Normal y = Normal (x + y) \land
    Normal v_0 + -\infty = -\infty \wedge \text{Normal } v_0 + +\infty = +\infty \wedge
    -\infty + Normal v_1 = -\infty \wedge +\infty + Normal v_1 = +\infty \wedge
    -\infty + -\infty = -\infty \wedge +\infty + +\infty = +\infty
[extreal_ainv_def]
\vdash --\infty = +\infty \land -+\infty = -\infty \land \forall x. -Normal x = \text{Normal } (-x)
[extreal sub]
\vdash x - y = x + -y
[extreal div]
\vdash (\forall r. Normal r / +\infty = Normal 0) \land
    (\forall r. \text{ Normal } r / -\infty = \text{Normal } 0) \land
    \forall x \ r. \ r \neq 0 \Rightarrow x \ / \text{ Normal } r = x \times (\text{Normal } r)^{-1}
[extreal_inv_def]
\vdash -\infty^{-1} = \text{Normal } 0 \land +\infty^{-1} = \text{Normal } 0 \land
   \forall r. \ r \neq 0 \Rightarrow (Normal \ r)^{-1} = Normal \ r^{-1}
```

NOTE: $\infty + -\infty$, $\infty - \infty$, $\frac{\infty}{\infty}$ and division-by-zero are unspecified.



Extended Real Numbers (2)

However, $0 \cdot \pm \infty = 0$ is defined to be 0 (consider a limiting process):

```
\vdash -\infty \times -\infty = +\infty \land -\infty \times +\infty = -\infty \land +\infty \times -\infty = -\infty \land +\infty \times +\infty
    Normal x \times -\infty =
    (if x = 0 then Normal 0
     else if 0 < x then -\infty
     else +\infty) \wedge
    -\infty \times \text{Normal } y =
    (if y = 0 then Normal 0
     else if 0 < y then -\infty
     else +\infty) \wedge
    Normal x \times +\infty =
    (if x = 0 then Normal 0
     else if 0 < x then +\infty
     else -\infty) \wedge
    +\infty \times \text{Normal } y =
    (if y = 0 then Normal 0
     else if 0 < y then +\infty
     else -\infty) \wedge Normal x \times Normal y = Normal (x \times y) [extreal_mul_def]
 \vdash 0 \times x = 0
                                                                                         [mul_lzero]
 \vdash x \times 0 = 0
                                                                                         [mul_rzero]
```



Measure Space

```
A measure space is a tuple (X, \mathcal{A}, \mu) (space, measurable sets, measure):
\square (X, \mathcal{A}) (measurable space) is a \sigma-algebra;
\mu: \mathcal{A} \to [0, +\infty] is a measure such that (cf. premeasure)
     \mu(\emptyset) = 0
     \forall s \in \mathcal{A}. \ 0 \leqslant \mu(s)
     \vdash measure_space m \iff
   sigma_algebra (measurable_space m) \land
   positive m \wedge \text{countably\_additive } m
                                                             [measure_space_def]
 \vdash positive m \iff
   measure m \emptyset = 0 \land
   \forall s. \ s \in \texttt{measurable\_sets} \ m \Rightarrow 0 < \texttt{measure} \ m \ s
                                                                   [positive_def]
 \vdash countably_additive m \iff
   \forall f. f \in (\mathcal{U}(:\text{num}) \rightarrow \text{measurable\_sets } m) \land
        (\forall i \ j. \ i \neq j \Rightarrow DISJOINT (f \ i) (f \ j)) \land
        \bigcup (IMAGE f \mathcal{U}(:num)) \in measurable\_sets m \Rightarrow
        measure m ([] (IMAGE f \mathcal{U}(:num))) =
        suminf (measure m \circ f)
                                                       [countably_additive_def]
```



More Properties of (pre)measure

```
\vdash increasing m \iff
  \forall s \ t.
      s \in \texttt{measurable\_sets} \ m \ \land
      t \in \texttt{measurable\_sets} \ m \ \land \ s \subseteq t \Rightarrow
      measure m s < measure m t
\vdash additive m \iff
  \forall s t.
      s \in \texttt{measurable\_sets} \ m \ \land
      t \in \texttt{measurable\_sets} \ m \ \land \ \texttt{DISJOINT} \ s \ t \ \land
      s \cup t \in \mathtt{measurable\_sets} \ m \Rightarrow
      measure m (s \cup t) = measure m s + measure m t
\vdash subadditive m \iff
  \forall s t
      s \in \texttt{measurable\_sets} \ m \ \land
      t \in \texttt{measurable\_sets} \ m \ \land
      s \cup t \in \texttt{measurable\_sets} \ m \Rightarrow
      measure m (s \cup t) < measure m s + measure m t
\vdash countably_subadditive m \iff
   \forall f. f \in (\mathcal{U}(:num) \rightarrow measurable\_sets m) \land
         [] (IMAGE f \ \mathcal{U}(:num)) \in measurable_sets m \Rightarrow
         measure m ([] (IMAGE f \mathcal{U}(:num))) \leq
         suminf (measure m \circ f)
```



Monotone Convergence of (pre)measure

$$f_n \subseteq f_{n+1} \in \mathcal{A} \Rightarrow \sup\{\mu(f_i)\} = \mu(\bigcup_{i \in \mathbb{N}} f_i)$$

[MONOTONE_CONVERGENCE2]

$$f_{n+1} \subseteq f_n \in \mathcal{A} \Rightarrow \inf\{\mu(f_i)\} = \mu(\bigcap_{i \in \mathbb{N}} f_i) \qquad (\mu(f_i) \neq \infty)$$

[MONOTONE_CONVERGENCE_BIGINTER2]

```
\vdash measure_space m \land f \in (\mathcal{U}(:\text{num}) \to \text{measurable\_sets } m) \land (\forall n. \text{ measure } m \ (f \ n) \neq +\infty) \land (\forall n. f \ (\text{SUC } n) \subseteq f \ n) \Rightarrow \text{inf } (\text{IMAGE } (\text{measure } m \circ f) \ \mathcal{U}(:\text{num})) = \text{measure } m \ (\bigcap \ (\text{IMAGE } f \ \mathcal{U}(:\text{num})))
```



Probability Space

A measure space (X, \mathcal{A}, μ) is called *probability space* if $\mu(X) = 1$.

```
\vdash prob_space p \iff measure_space p \land measure p \pmod{p} = 1 [prob_space_def]
```

Probability space is usually denoted by (Ω, \mathcal{A}, P) , where:

- Ω is called the sample space;
- \square A is called the set of events and $E \in \mathcal{A}$ a event;
- ightharpoonup P(E) is called the *probability* of event E.

```
\vdash prob_space p\iff sigma_algebra (p_space p, events p) \land positive p \land countably_additive p \land prob p (p_space p) = 1 [PROB_SPACE]
```

A sample probability space (for one-time coin tossing): $(\{H, T\}, \{\emptyset, \{H\}, \{T\}, \{H, T\}\}, P)$, where $P\{H\} = P\{T\} = 0.5$.



Independent Events; Conditional Probability

Two events A and B are independent if $P(A \cap B) = P(A) P(B)$:

$$\begin{array}{c} \vdash \text{ indep } p \text{ } a \text{ } b \iff \\ a \in \text{ events } p \wedge b \in \text{ events } p \wedge \\ \text{ prob } p \text{ } (a \cap b) = \text{ prob } p \text{ } a \times \text{ prob } p \text{ } b \end{array} \qquad \text{[indep_def]} \\ P(E_1|E_2) := \frac{P(E_1 \cap E_2)}{P(E_2)} \quad \text{(conditional probability)} \end{array}$$

$$\vdash$$
 cond_prob p e_1 e_2 = prob p $(e_1 \cap e_2)$ / prob p e_2

[cond_prob_def]

[BAYES_RULE]

Bayes' formula:

$$P(B|A) = P(A|B) \cdot \frac{P(B)}{P(A)}$$
 $(P(A) \neq 0, P(B) \neq 0)$

$$\vdash$$
 prob_space $p \land A \in$ events $p \land B \in$ events $p \land$ prob $p \land A \neq 0 \land$ prob $p \land B \neq 0 \Rightarrow$ cond_prob $p \land A \Rightarrow$ cond_prob $p \land A \Rightarrow$ prob $p \land A \Rightarrow$ prob $p \land A \Rightarrow$ prob $p \land A \Rightarrow$

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Exercises

- 1. Prove $(X, \{\emptyset, X\})$ is $(\sigma$ -)algebra.
- 2. Prove $(\{H, T\}, \{\emptyset, \{H\}, \{T\}, \{H, T\}\})$ is $(\sigma$ -)algebra.
- 3. Prove $(\{H,T\},\{\emptyset,\{H\},\{T\},\{H,T\}\},P)$, where $P\{H\}=P\{T\}=0.5$, is indeed a probability space.

Some useful theorems:

```
[sigma_algebraTheory.algebra_finite_space_imp_sigma_algebra] \vdash algebra a \land FINITE (space a) \Rightarrow sigma_algebra a [measureTheory.finite_additivity_sufficient_for_finite_spaces2] \vdash sigma_algebra (measurable_space m) \land FINITE (m_space m) \land positive m \land additive m \Rightarrow measure_space m
```

