

Knuth-Skilling Formalization

Review Walkthrough

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Abstract

This document provides a systematic walkthrough of the Lean 4 formalization of Knuth & Skilling's "Foundations of Inference" [1]. Each section lists the main theorem statements with their Lean proofs and file locations, organized by the corresponding K&S paper sections. The reviewer entrypoint `.../KnuthSkilling/FoundationsOfInference.lean` is **complete with zero sorries**. Supplementary modules (Cox [9], Shore–Johnson [10], Shannon [7]/Faddeev [12]) are present but intentionally not imported by default by the FOI reviewer entrypoint; they provide alternative derivation paths and may assume additional regularity gates.

Two-track reading (choose your level). This walkthrough is structured so that a mathematician can audit *statements* quickly, while a proof-assistant user can audit *proofs* precisely.

Track A: theorem-level audit (skim).

- Start with Section 13 (Summary), then jump to any topic and read only the definitions + theorem statements.
- Use file/line pointers for rapid spot-checking without reading every proof.

Track B: proof-level audit (deep).

- Run `lake build Mettapedia.ProbabilityTheory.KnuthSkilling.FoundationsOfInference`.
- Open the referenced Lean files at the cited line numbers to audit the full proof scripts.

Scope note. The reviewer entrypoint `FoundationsOfInference.lean` intentionally excludes optional derivation paths (Cox, Shore–Johnson, Shannon/Faddeev, algorithmic probability), which are present as opt-in modules.

Quickstart for Reviewers

- **High-level view:** Start with Section 13 (Summary) for the consolidated theorem list and key discoveries.
- **Audit Appendix A (sum rule):** See Section 5, then the three proof paths (Hölder / Grid / Direct Cuts).

*This document was drafted collaboratively by humans and AI systems (GPT/Codex, Claude). While formal claims are machine-checked in Lean 4, prose descriptions may contain human or AI hallucinations. Caveat lector.

- **Audit Appendix B (product rule):** See Section 6 for the two independent proof paths.
- **Audit Appendix C (entropy / variational theorem):** See Section 7 and the counterexamples in Section 12.
- **Audit σ -additivity bridge (extension):** See Section 10 (`Core/ScaleCompleteness.lean`).
- **Build commands:** See Section 14 (Build Instructions) at the end.

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1 Overview and File Structure

1.1 K&S Paper Coverage

K&S Section	Topic	Lean Files
Sections 1–2	Sum-side Axioms (Sym 0–2)	Core/Basic.lean, Core/Algebra.lean
Sections 3,7	Probability (two K&S paths)	Probability/ProbabilityDerivation.lean, Probability/ConditionalProbability/Basic.lean
Section 4	Quantum theory	Core/SymmetricalFoundation.lean, Mettapedia/Algebra/TwoDimClassification.lean
Section 6	Divergence	Information/Divergence.lean
Section 8	Info/Entropy	Information/InformationEntropy.lean
Appendix A	Representation	Additive/Main.lean, Additive/Representation.lean
Appendix B	Product (Sym 3–4)	Multiplicative/Main.lean, Multiplicative/ScaledMultRep.lean
Appendix C	Variational	Variational/Main.lean
Extension	σ -additivity bridge	Core/ScaleCompleteness.lean

1.2 Key Files

All paths below are relative to `Mettapedia/ProbabilityTheory/KnuthSkilling/`.

`./KnuthSkilling.lean` Main entrypoint (FOI core + extra K&S modules)

`FoundationsOfInference.lean` Reviewer entrypoint (FOI core; no WIP)

`Core/Basic.lean` Core axioms: `KSSemigroupBase`, `KnuthSkillingMonoidBase`, `KnuthSkillingAlgebraBase`

`Core/Algebra.lean` `iterate_op`, separation axioms (`KSSeparation*`)

`Core/ScaleCompleteness.lean` σ -completeness axioms and σ -additivity theorem

`Core/SymmetricalFoundation.lean` Section 4 quantum derivation

`Additive/Main.lean` Appendix A entrypoint (typeclass interface + instances)

`Additive/Representation.lean` Appendix A representation interfaces (identity-free default)

`Additive/Axioms/SandwichSeparation.lean` Archimedean + commutativity from `KSSeparation`

`Additive/Proofs/OrderedSemigroupEmbedding/HolderEmbedding.lean` Hölder/Alimov embedding path

`Additive/Proofs/GridInduction/Main.lean` Grid/induction path (K&S-style globalization)

Additive/Proofs/DirectCuts/Main.lean Dedekind cuts path (alternative)

Multiplicative/Main.lean Appendix B pipeline (product \Rightarrow exponential \Rightarrow scaled mult)

Multiplicative/Proofs/Direct/DirectProof.lean Appendix B direct algebraic proof path

Multiplicative/ScaledMultRep.lean Appendix B common interface for both proof paths

Variational/Main.lean Appendix C variational theorem (entropy form)

Information/Divergence.lean Section 6 divergence

Information/InformationEntropy.lean Section 8 information/entropy (KL, Shannon)

Probability/ConditionalProbability/Basic.lean Section 7 conditional probability (lattice path)

2 Core Sum-Side Axioms (K&S Sections 1–2)

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Core/Basic.lean

2.1 K&S Sum-Side Symmetries (0–2)

K&S present symmetries in two groups. The **sum-side** symmetries (0–2) govern the combination operation \oplus :

- **Symmetry 0** (Fidelity): $\bar{x} < \bar{y} \Rightarrow x < y$
- **Symmetry 1** (Monotonicity): $\bar{x} < \bar{y} \Rightarrow \bar{x} \oplus \bar{z} < \bar{y} \oplus \bar{z}$
- **Symmetry 2** (Associativity): $(\bar{x} \oplus \bar{y}) \oplus \bar{z} = \bar{x} \oplus (\bar{y} \oplus \bar{z})$

The **product-side** symmetries (3–4) are formalized separately in Multiplicative/Main.lean (see Section 5).

Definition 2.1 (KSsemigroupBase). Lines 180–188, Core/Basic.lean

Identity-free core structure containing only Symmetries 0–2.

Listing 1: Identity-free semigroup base

```
1 class KSsemigroupBase (alpha : Type*) extends LinearOrder alpha where
2   op : alpha → alpha → alpha           -- combination operation
3   op_assoc : ∀ x y z : alpha, op (op x y) z = op x (op y z)  -- Sym 2
4   op_strictMono_left : ∀ y : alpha, StrictMono (fun x => op x y)  -- Sym 0+1
5   op_strictMono_right : ∀ x : alpha, StrictMono (fun y => op x y)  -- Sym 0+1
```

Definition 2.2 (KnuthSkillingMonoidBase). Lines 237–244, Core/Basic.lean

Adds an identity element (`ident`) without assuming it is the order minimum.

Listing 2: Core K&S structure with identity (no positivity assumption)

```
1 class KnuthSkillingMonoidBase (alpha : Type*) extends KSsemigroupBase alpha where
2   ident : alpha
3   op_ident_right : ∀ x : alpha, op x ident = x
4   op_ident_left : ∀ x : alpha, op ident x = x
```

Definition 2.3 (KnuthSkillingAlgebraBase). Lines 246–248, Core/Basic.lean

The probability-theory convenience layer: assumes the identity is the order minimum (`ident_le`).

Listing 3: K&S probability base: identity is minimum (positivity)

```

1 class KnuthSkillingAlgebraBase (alpha : Type*) extends KnuthSkillingMonoidBase
2   alpha where
3     ident_le : ∀ x : alpha, ident ≤ x

```

Remark 2.4 (Implicit Linear Order). K&S never explicitly state that elements are totally ordered, but their proofs rely on trichotomy. We make this explicit via `LinearOrder`.

Remark 2.5 (Identity Element—Essential for Positivity). **Important:** The identity element (`ident`) is **not** among K&S’s numbered symmetries. K&S explicitly state that the bottom element \perp is **optional**:

“with the bottom element optional” (K&S line 320)

“Some mathematicians opt to include the bottom element on aesthetic grounds, whereas others opt to exclude it” (K&S lines 340–341)

However, our formalization **disproves** K&S’s claim that fidelity alone ensures positivity. The \mathbb{Z} counterexample (`Additive/Counterexamples/NegativeWithoutIdentity.lean`) shows:

- $(\mathbb{Z}, +, \leq)$ satisfies K&S Axioms 1–2 but has no identity
- The representation theorem applies, giving $\Theta : \mathbb{Z} \rightarrow \mathbb{R}$
- But $\Theta(-1) = -1 < 0$ —**negative values appear**

Corrected understanding:

- **With identity + ident_le:** $\Theta(\perp) = 0$; all $x > \perp$ have $\Theta(x) > 0$
- **Without identity:** Representation works but positivity is **not guaranteed**

Identity with `ident_le` (i.e., $\perp \leq x$ for all x) is **essential** for positivity, not “aesthetic.”

Listing 4: \mathbb{Z} counterexample: representation works but positivity fails

```

1 -- Z is a valid KSSemigroupBase (satisfies Axioms 1-2)
2 instance Int.instKSSemigroupBase : KSSemigroupBase Int where
3   op := (. + .)
4   op_assoc := add_assoc
5   op_strictMono_left := fun y => by intro a b hab; omega
6   op_strictMono_right := fun x => by intro a b hab; omega
7
8 -- But Z cannot satisfy ident_le (no minimum element)
9 theorem Int.cannot_satisfy_ident_le : ¬(∀ n : Int, (0 : Int) ≤ n) := by
10   push_neg; use -1; omega
11
12 -- The Holder embedding has negative values: Phi(-1) < 0
13 theorem Int.holder_embedding_has_negatives :
14   ∃ (G : Subsemigroup (Multiplicative R))
15     (Theta : Multiplicative Int ≈o G),
16     Multiplicative.toAdd (Theta (Multiplicative.ofAdd (-1))) < 0 :=
17     holder_embedding_produces_negatives MultiplicativeInt.no_anomalous_pair

```

Remark 2.6 (Unbundled Axiom Predicates). **Lines 142–164, Core/Basic.lean**

In addition to the bundled typeclasses, we provide **unbundled predicates** for each axiom. This enables flexible hypothesis tracking—use individual predicates when you need minimal assumptions, or bundled classes when you want ergonomic access to multiple axioms.

Sum-side predicates (Core/Basic.lean):

- `OpAssoc op` (line 143): $\forall x y z, \text{op}(\text{op}(x, y), z) = \text{op}(x, \text{op}(y, z))$
- `OpStrictMonoLeft op` (line 147): $\forall y, \text{StrictMono}(\lambda x. \text{op}(x, y))$
- `OpStrictMonoRight op` (line 151): $\forall x, \text{StrictMono}(\lambda y. \text{op}(x, y))$
- `OpIdentLeft op e` (line 155): $\forall x, \text{op}(e, x) = x$
- `OpIdentRight op e` (line 159): $\forall x, \text{op}(x, e) = x$
- `IdentIsMin e` (line 163): $\forall x, e \leq x$

Connection theorems: The `KSSemigroupBase` and `KnuthSkillingAlgebraBase` namespaces provide lemmas like `KSSemigroupBase.opAssoc` that extract the unbundled predicate from a bundled instance.

2.2 Iteration

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Core/Algebra.lean

Definition 2.7 (iterate_op). **Lines 23–25, Algebra.lean**

K&S Paper Reference: This corresponds to K&S’s use of “ n copies of x ” in their proofs, written as x^n or nx depending on context (see K&S equations around line 370–380 in the TeX source). The iteration builds repeated applications of \oplus : $x^n = \underbrace{x \oplus x \oplus \cdots \oplus x}_{n \text{ times}}$.

Listing 5: Iteration definition

```

1 def iterate_op (x : alpha) : N → alpha
2 | 0 => ident
3 | n + 1 => op x (iterate_op x n)

```

This builds the sequence: $\text{ident}, x, x \oplus x, x \oplus (x \oplus x), \dots$

2.3 Identity-Free Iteration

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Core/Basic.lean

Definition 2.8 (iterate_op_pnat). **Lines 375–376, Core/Basic.lean**

For identity-free reasoning, we define iteration using positive natural numbers (\mathbb{N}^+) instead of \mathbb{N} . This works on the weaker `KSSemigroupBase` (no identity required).

Listing 6: Identity-free iteration

```

1 def iterate_op_pnat [KSSemigroupBase alpha] (x : alpha) (n : N+) : alpha :=
2   iterate_op_pnat_aux x (n.val - 1)
3
4 private def iterate_op_pnat_aux (x : alpha) : N → alpha
5   | 0 => x           -- n=0 maps to  $x^1 = x$ 
6   | n + 1 => op x (iterate_op_pnat_aux x n)

```

Remark 2.9 (Key Properties). • `iterate_op_pnat x 1 = x` (base case is x , not ident)

- `iterate_op_pnat x (n+1) = op x (iterate_op_pnat x n)` (recursion)
- Builds the sequence: $x, x \oplus x, x \oplus (x \oplus x), \dots$ (no identity!)
- **Connection:** `iterate_op_pnat x n = iterate_op x n.val` when $n \geq 1$ (Core/Algebra.lean:198)

Usage: This identity-free version is used in:

- `KSSeparationSemigroup / KSSeparationSemigroupStrict` (Core/Algebra.lean)
- `Additive/Proofs/OrderedSemigroupEmbedding/HolderEmbedding.lean` (Hölder path)
- `Additive/Proofs/DirectCuts/DirectCuts.lean` (cuts path)

3 No Anomalous Pairs and Separation

This section addresses the critical question: *what additional property, beyond order and associativity, is required to guarantee an additive embedding into $(\mathbb{R}, +)$?*

Knuth & Skilling's Appendix A proof is constructive, building a grid of values without assuming continuity. The formalization must make explicit certain density requirements that their constructive approach handles implicitly. Two roughly equivalent formulations emerge:

- **No Anomalous Pairs (NAP):** The classical condition from ordered semigroup theory [2, 3, 4].
- **Separation:** A “sandwich” property extracted from K&S’s constructive proof, made explicit in the formalization.

The formalization uses NAP as the *primary* axiom because it connects to 60+ years of classical algebra and has a complete machine-checked proof via Eric Luap’s `OrderedSemigroups` library.

3.1 No Anomalous Pairs (Classical Formulation)

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Additive/Axioms/AnomalousPairs.lean

Definition 3.1 (Anomalous Pair). Two elements a, b of an ordered semigroup form an *anomalous pair* if their iterates remain “squeezed” forever:

$$a^n < b^n < a^{n+1} \quad \text{for all } n \in \mathbb{N}^+$$

(or the symmetric condition $a^n > b^n > a^{n+1}$ for negative elements).

Definition 3.2 (NoAnomalousPairs). An ordered semigroup has *no anomalous pairs* if no such pair exists.

Listing 7: NAP definition (following Eric Luap’s OrderedSemigroups)

```

1 def AnomalousPair (a b : alpha) : Prop :=
2   ∀ n : N+,
3     (iterate_op_pnat a n < iterate_op_pnat b n) ∧
4       iterate_op_pnat b n < iterate_op_pnat a (n + 1)) ∨/

```

```

5   (iterate_op_pnat a n > iterate_op_pnat b n /\ 
6     iterate_op_pnat b n > iterate_op_pnat a (n + 1))
7
8 class NoAnomalousPairs (alpha : Type*) [KSSemigroupBase alpha] : Prop where
9   not_anomalous :  $\forall$  a b : alpha, Not (AnomalousPair a b)

```

Historical context: For *groups*, Hölder [2] proved that Archimedean ordered groups embed into $(\mathbb{R}, +)$. For *semigroups* (without inverses), Alimov [3] identified “no anomalous pairs” as the precise generalization. Fuchs [4] provided the textbook treatment.

Theorem 3.3 (Hölder–Alimov [2, 3, 4]). *In a linearly ordered cancellative semigroup, the following are equivalent:*

1. *The semigroup has no anomalous pairs.*
2. *There exists an order-preserving additive embedding into $(\mathbb{R}, +)$.*

This classical result is formalized in Eric Luap’s `OrderedSemigroups` library [5] and imported into the K&S development.

3.2 Separation (K&S Constructive Formulation)

File: `Mettapedia/ProbabilityTheory/KnuthSkilling/Core/Algebra.lean`

Knuth & Skilling’s Appendix A proof works constructively: they build a grid of values by introducing atoms one at a time, placing each new value at a “convenient” point within an interval. This implicitly assumes a *density* property—that for any interval (x, y) , powers of a base element can be placed within it.

The formalization makes this explicit as the **separation property**:

Definition 3.4 (`KSSeparationSemigroup`). For any positive elements a, x, y with $x < y$, there exist exponents $(n, m) \in \mathbb{N}^+$ such that:

$$x^m < a^n \leq y^m$$

Listing 8: Separation axiom (identity-free)

```

1 class KSSeparationSemigroup (alpha : Type*) [KSSemigroupBase alpha] where
2   separation :  $\forall$  {a x y : alpha},
3     IsPositive a  $\rightarrow$  IsPositive x  $\rightarrow$  IsPositive y  $\rightarrow$  x < y  $\rightarrow$ 
4      $\exists$  n m : N+, iterate_op_pnat x m < iterate_op_pnat a n /\ 
5       iterate_op_pnat a n  $\leq$  iterate_op_pnat y m

```

Intuition: The separation property says that powers of any positive base element are “dense enough” to separate any two distinct elements. This is precisely the density that K&S’s constructive proof requires to place new grid points.

Key advantage: The formulation works on `KSSemigroupBase` without requiring identity, following Eric Luap’s approach.

3.3 Equivalence Theorem

File: `Mettapedia/ProbabilityTheory/KnuthSkilling/Additive/Axioms/AnomalousPairs.lean`

Under our standing hypotheses, the two formulations are equivalent:

Theorem 3.5 (Equivalence of NAP and Separation). *For a linearly ordered cancellative semigroup with strictly monotone operation and identity as minimum:*

$$\text{KSSeparation} \iff \text{NoAnomalousPairs} \iff \text{Additive real representation}$$

Proof sketch:

1. **Separation \Rightarrow NAP:** If (a, b) were anomalous with $a^n < b^n < a^{n+1}$ for all n , separation would provide witnesses (n, m) with $a^m < c^n \leq b^m$ for some base c —breaking the squeeze. (`noAnomalousPairs_of_KSSeparation_with_IdentMin`)
2. **NAP \Rightarrow Embedding:** The Hölder/Alimov theorem (Theorem 3.3), formalized via `OrderedSemigroups.holder_not_anom`.
3. **Embedding \Rightarrow Separation:** Rational density in \mathbb{R} provides the sandwich witnesses. For any $x < y$ in the semigroup, their images $\Theta(x) < \Theta(y)$ in \mathbb{R} have a rational p/q between them, which translates back to the required power witnesses.

3.3.1 Bilateral Separation

Files:

- `Core/Algebra.lean` — `SeparationGeneralizedProp` definition
- `Additive/Proofs/GridInduction/Comparison.lean` — proofs

Without the assumption that identity is a minimum, elements can exist *below* identity (mapping to negative reals under Θ). The separation axiom extends naturally to a **bilateral** form:

Definition 3.6 (`SeparationGeneralizedProp`). *Lines 400–409, Core/Algebra.lean*
Bilateral separation consists of two branches:

- **(Positive)** For `ident < a` and `ident < x < y`: sandwich condition.
- **(Negative)** For `a < ident` and `x < y < ident`: same sandwich condition.

The key constraint: a, x, y must lie on the *same side* of identity.

Theorem 3.7 (`separationGeneralizedProp_of_noAnomalousPairs`). *Lines 563–570, Additive/Proofs/GridInduction/Comparison.lean*

NoAnomalousPairs implies SeparationGeneralizedProp.

Proof idea (negative branch): From NAP, obtain $\Theta : \alpha \rightarrow \mathbb{R}$. For $a, x, y < \text{ident}$, we have $\Theta(a), \Theta(x), \Theta(y) < 0$. Setting $A = -\Theta(a)$, $X = -\Theta(y)$, $Y = -\Theta(x)$ (note the swap!) gives $0 < X < Y$. By Archimedean property, pick m with $A \leq m(Y - X)$, then $n := \lceil mX/A \rceil$ gives $mX \leq nA < mY$. Translating back yields the sandwich $x^{(m)} < a^{(n)} \leq y^{(m)}$.

Remark 3.8 (Probability setting). Under `IdentIsMinimum` (the probability setting), no elements exist below identity, so the negative branch is vacuously true and `SeparationGeneralizedProp` reduces to the standard one-sided `KSSeparation`.

Remark 3.9 (Why NAP is Primary). The formalization treats `NoAnomalousPairs` as the primary axiom for three reasons:

1. **Historical precedent:** NAP predates K&S by 60+ years [2].

2. **Verified library:** Eric Luap's `OrderedSemigroups` provides a complete, machine-checked proof of the Hölder embedding theorem.
3. **Minimality:** NAP is identity-free and works at the semigroup level, making it the most general formulation.

3.4 Derived Properties

From NAP (equivalently, Separation), we derive two key properties that K&S claim follow from order + associativity alone:

3.4.1 Archimedean Property

File: `Additive/Axioms/SandwichSeparation.lean`

Theorem 3.10 (archimedean_of_separation_pos). *Under `KSSeparation`, for any positive a and x , there exists n such that $x \leq a^n$.*

This confirms there are no “infinitesimals”—every element is eventually exceeded by powers of any positive base.

3.4.2 Commutativity

File: `Additive/Axioms/SandwichSeparation.lean`

Theorem 3.11 (ksSeparation_implies_comm). *Under `KSSeparation`, the operation is commutative: $x \oplus y = y \oplus x$.*

K&S (lines 1160–1163 of their paper): “Commutativity was not assumed either, though commutativity of the resulting measure follows as a property of additivity.”

Two proof routes:

1. **Via Hölder embedding:** NAP \Rightarrow embeds into $(\mathbb{R}, +)$ \Rightarrow commutativity (since \mathbb{R} is commutative).
2. **Direct mass counting:** Assume $x \oplus y < y \oplus x$. Apply separation to get $(x \oplus y)^m < a^n \leq (y \oplus x)^m$. By associativity, $(x \oplus y)^n$ and $(y \oplus x)^m$ contain the same “atom counts,” leading to contradiction when $n > m$.

Both proofs are formalized; the Hölder route is canonical.

Remark 3.12 (K&S’s Insight). K&S’s key observation (lines 1156–1163):

“We find that associativity and order provide minimal assumptions that are convincing and compelling for scalar additivity in all its applications. Associativity alone does not force additivity, but associativity with order does.”

The formalization confirms this, but makes explicit that “order” must include the density/separation requirement—not just strict monotonicity. This is not a gap in K&S’s argument; it is implicit in their constructive proof, which assumes the ability to place new values within intervals.

4 The Representation Theorem (Appendix A)

Files:

- `Additive/Representation.lean` (interfaces: identity-free default)
- `Additive/Main.lean` (entrypoint: lightweight interface + instances)
- `Additive/Proofs/OrderedSemigroupEmbedding/HölderEmbedding.lean` (Hölder/Alimov embedding path)
- `Additive/Proofs/DirectCuts/Main.lean` (Dedekind cuts path)
- `Additive/Proofs/GridInduction/Main.lean` (Grid/induction path; K&S-style, large)
- `Additive/Proofs/GridInduction/Globalization.lean` (globalization orchestrator)
- `Additive/Proofs/GridInduction/Core/` ($\sim 17,700$ lines: grid infrastructure)
 - `Core/MultiGrid.lean`: AtomFamily, MultiGridRep, grid representations
 - `Core/Induction/`: Inductive extension theorems (Construction, ThetaPrime, DeltaShift)
 - `Core/OneDimensional.lean`: Base case (single atom)
 - `Core/Prelude.lean`: Foundational lemmas

Architecture: `Additive/Proofs/GridInduction/Globalization.lean` imports `Additive/Proofs/GridInduction/Core/All.lean` and orchestrates the grid machinery to prove global representation.

4.1 Three Independent Proof Paths

The formalization provides **three complete, independent proof routes** to the representation theorem:

1. **Hölder embedding** (MAIN ROUTE, weakest assumptions): Uses the NoAnomalousPairs condition and classical ordered semigroup theory [2, 3, 4]. Formalized via Eric Luap's `OrderedSemigroups` [5].
2. **Dedekind cuts** (alternative): Uses Separation property with Hölder/Dedekind cuts construction, bypassing the grid machinery.
3. **Grid induction** (K&S-style): Uses multi-dimensional grid representations and induction on atom families, following K&S's original approach.

Historical development:

- First: Grid/induction path (following K&S's original approach)
- Second: Cuts path discovered by Claude Code as superior
- Third: Hölder/Alimov path discovered by GPT-5.2 Pro as the best (weakest assumptions, classical connection)

All three proofs are complete with zero sorries. The Hölder path uses **NoAnomalousPairs only**—the weakest known condition sufficient for the representation theorem.

4.2 Proof Architecture 1: The Hölder Path (Main Route)

File: Additive/Proofs/OrderedSemigroupEmbedding/HölderEmbedding.lean

The Hölder path is the **recommended main route** because it uses the weakest hypotheses and connects to classical ordered semigroup theory.

Theorem 4.1 (holder_embedding_of_noAnomalousPairs). *Lines 166–169, HolderEmbedding.lean*

If a K&S algebra has no anomalous pairs, it embeds into \mathbb{R} .

Listing 9: Hölder embedding theorem

```

1 theorem holder_embedding_of_noAnomalousPairs [NoAnomalousPairs alpha] :
2   ∃ G : Subsemigroup (Multiplicative R), Nonempty (alpha =equiv)*o G) := by
3   have h : ~has_anomalous_pair (alpha := alpha) := noAnomalousPairs_iff_eric
4   exact holder_not_anom h

```

Theorem 4.2 (representation_semigroup). *Lines 272–278, HolderEmbedding.lean*

Identity-free representation: NoAnomalousPairs implies additive embedding into \mathbb{R} . Θ is defined up to an additive constant (no canonical zero point).

Listing 10: Identity-free representation theorem

```

1 theorem representation_semigroup [NoAnomalousPairs alpha] :
2   ∃ Theta : alpha → R,
3     (∀ a b : alpha, a ≤ b ↔ Theta a ≤ Theta b) ∧\
4     (∀ x y : alpha, Theta (op x y) = Theta x + Theta y) := by
5   obtain <G, <iso>> := holder_embedding_of_noAnomalousPairs (alpha := alpha)
6   use theta_from_embedding G iso
7   exact <theta_preserves_order G iso, theta_additive G iso>

```

Theorem 4.3 (representation_from_noAnomalousPairs). *Lines 300–307, HolderEmbedding.lean*

With identity: NoAnomalousPairs implies the full representation with $\Theta(\text{ident}) = 0$.

Listing 11: Representation with identity normalization

```

1 theorem representation_from_noAnomalousPairs [NoAnomalousPairs alpha] :
2   ∃ Theta : alpha → R,
3     (∀ a b : alpha, a ≤ b ↔ Theta a ≤ Theta b) ∧\
4     Theta ident = 0 ∧\
5     ∀ x y : alpha, Theta (op x y) = Theta x + Theta y := by
6   obtain <G, <iso>> := holder_embedding_of_noAnomalousPairs (alpha := alpha)
7   use theta_from_embedding G iso
8   exact <theta_preserves_order G iso, theta_ident G iso, theta_additive G iso>

```

Key advantage: Identity-free semigroup version:

representation_semigroup (for KSSemigroupBase).

With identity, the same path yields representation_from_noAnomalousPairs (for KnuthSkillingAlgebraBase).

4.3 Proof Architecture 2: The Grid/Induction Path

The grid-based proof is packaged as the typeclass `RepresentationGlobalization`, which is automatically instantiated when `[KSSeparationStrict α]` is available.

Definition 4.4 (RepresentationGlobalization). Lines 54–60, Globalization.lean

A typeclass packaging the existence of Θ .

Listing 12: `RepresentationGlobalization` typeclass

```

1 class RepresentationGlobalization (alpha : Type*)
2   [KnuthSkillingAlgebra alpha] [KSSeparation alpha] : Prop where
3   Θ : Type
4   Θ : α → R,
5   (forall a b : α, a ≤ b ↔ Θ a ≤ Θ b) ∧
6   Θ ident = 0 ∧
7   ∀ x y : α, Θ (op x y) = Θ x + Θ y

```

4.3.1 The Globalization Construction (“Triple Family Trick”)

The instance `representationGlobalization_of_KSSeparationStrict` (lines 93–850, `Globalization.lean`) constructs Θ globally using a multi-step process:

1. **Reference atom:** Choose any $a_0 > \text{ident}$ as a fixed reference point.
2. **2-atom families:** For each $x > \text{ident}$, build a 2-atom family $F_2 = \{a_0, x\}$ with a `MultiGridRep` R_2 (via `extend_grid_rep_with_atom_of_KSSeparationStrict` from `Core/`).
3. **Define $\Theta(x)$:** Extract the representation value from the grid:

$$\Theta(x) := R_2.\Theta_grid(\langle x, \text{membership_proof} \rangle)$$

4. **Well-definedness:** Use 3-atom families $F_3 = \{a_0, a_1, x\}$ to show that $\Theta(x)$ does not depend on the choice of reference atom. Path independence follows from `DeltaSpec_unique` (line 755, `Core/Induction/Construction.lean`).
5. **Order preservation:** For $a < b$, build $F_3 = \{a_0, a, b\}$ and use `MultiGridRep.strictMono` to show $\Theta(a) < \Theta(b)$.
6. **Additivity:** For $x \oplus y$, build $F_3 = \{a_0, x, y\}$ and verify $\Theta(x \oplus y) = \Theta(x) + \Theta(y)$ by path independence across different extension orderings.

Remark 4.5 (Why “Triple Family Trick”?). The name comes from using 3-atom families to mediate between different 2-atom constructions. This technique ensures global consistency: any two definitions of $\Theta(x)$ via different reference atoms must agree, because they both embed into a common 3-atom grid representation.

Remark 4.6 (Identity-Free Grid Infrastructure). The grid construction has **parametric versions** that could work without identity:

- `mu_param F r base`: Grid valuation with explicit base element instead of `ident`
- `kGrid_param F base`: Grid set using `mu_param`

- `mu_pnat, kGrid_pnat`: Truly identity-free using \mathbb{N}^+ iteration (no 0 exponents)
- `RepresentationGlobalizationAnchor`: Class for representations normalizing to an arbitrary anchor

Currently, the globalization instance uses identity:

`representationGlobalization_of_KSSeparationStrict`. An identity-free instance using the parametric infrastructure is marked as **future work** in `Globalization.lean`.

For identity-free representations `today`, use the Hölder path (`HolderEmbedding.lean`) which produces `RepresentationResult` (order + additivity, no normalization constraint).

4.4 Main Theorem Statement

Theorem 4.7 (associativity_representation). *Lines 54–60, Additive/Proofs/GridInduction/Main.lean*

K&S Appendix A Main Theorem: *There exists an order embedding $\Theta : \alpha \rightarrow \mathbb{R}$ such that:*

1. *Order preservation:* $a \leq b \Leftrightarrow \Theta(a) \leq \Theta(b)$
2. $\Theta(\text{ident}) = 0$
3. *Additivity:* $\Theta(op x y) = \Theta(x) + \Theta(y)$

Listing 13: Appendix A Representation Theorem (public API)

```

1 theorem associativity_representation
2   (alpha : Type*) [KnuthSkillingMonoidBase alpha] [KSSeparation alpha]
3   [RepresentationGlobalization alpha] :
4   ∃ Theta : alpha → ℝ,
5   ( ∀ a b : alpha, a ≤ b ↔ Theta a ≤ Theta b ) ∧
6   Theta ident = 0 ∧
7   ∀ x y : alpha, Theta (op x y) = Theta x + Theta y := by
8   exact RepresentationGlobalization.∃_Theta (alpha := alpha)

```

Remark 4.8 (Proof Delegation). The theorem statement simply extracts `exists_Theta` from the typeclass. All the actual work happens in the instance construction:

`representationGlobalization_of_KSSeparationStrict`
(starts at line 105, `Additive/Proofs/GridInduction/Globalization.lean`)

This design keeps the public API clean while hiding the complex globalization machinery.

4.5 Proof Architecture 2: The Direct Cuts Path

File: `Additive/Proofs/DirectCuts/DirectCuts.lean`

The DirectCuts path provides **both identity-based and identity-free** versions using Dedekind cuts:

- **Identity-free:** Uses `Theta_cuts_pnat` with \mathbb{N}^+ iteration
 - `Theta_cuts_pnat` (line 1434): Definition via Dedekind cuts using \mathbb{N}^+ iteration
 - `Theta_cuts_pnat_strictMono` (line 1530): Strict monotonicity (fully proven)
 - `Theta_cuts_pnat_add` (line 1636): Additivity (fully proven)
 - No reference to `ident` anywhere

- **Identity-based** (§9a): Uses `Theta_cuts` with \mathbb{N} iteration
 - `iterate_op x 0 = ident` for the base case
 - `ident` as the canonical reference point
 - Produces `RepresentationResult` satisfying $\Theta(\text{ident}) = 0$

The cuts construction uses a classical Hölder/Dedekind approach (shown here for the identity-based version; the identity-free version uses `IsPositive` instead of comparing to `ident`):

1. **Fix base element:** Choose any $a_0 > \text{ident}$ as a reference point (identity-free: choose any a_0 with `IsPositive a0`)
2. **Define rational approximants:** For any $x \in \alpha$, consider the set of ratios $m/n \in \mathbb{Q}$ where $a_0^m \leq x^n$ (equivalently, $m \cdot a_0 \leq n \cdot x$ in additive notation)
3. **Define $\Theta(x)$ by supremum in \mathbb{R} :**

$$\Theta_{\text{cuts}}(x) := \sup_{\mathbb{R}} \{m/n \in \mathbb{Q} : a_0^m \leq x^n, n > 0\}$$

where the supremum is taken in \mathbb{R} (which is already complete from Mathlib). The cut set is defined in α using the order relation, but the supremum is computed in \mathbb{R} .

4. Prove properties:

- **Order preservation:** If $x < y$, then for any m/n in the cut of x , there exists m'/n' in the cut of y with $m/n < m'/n'$ (uses `KSSeparation` to find witnesses)
- **Additivity:** $\Theta(x \oplus y) = \Theta(x) + \Theta(y)$ follows from $a_0^{m_1+m_2} \leq (x \oplus y)^{n_1+n_2}$ iff $a_0^{m_1} \leq x^{n_1}$ and $a_0^{m_2} \leq y^{n_2}$ (uses commutativity and associativity)

Remark 4.9 (No Circularity). This construction does **not** require completing α into \mathbb{R} first. Instead:

- The set $\{m/n \in \mathbb{Q} : a_0^m \leq x^n\}$ is defined using the order relation in α
- These rationals are cast to \mathbb{R} : `(↑) ``cutSet a x : Set ℝ`
- The supremum is computed in \mathbb{R} using `sSup` (conditional supremum from Mathlib)

Thus $\Theta : \alpha \rightarrow \mathbb{R}$ is directly defined without requiring α to already embed into \mathbb{R} .

Theorem 4.10 (associativity_representation_cuts). *Lines 44–71, Additive/Proofs/DirectCuts/Main.lean*

The cuts-based representation theorem.

Listing 14: Appendix A (cuts proof)

```

1 theorem associativity_representation_cuts
2   (alpha : Type*) [KnuthSkillingAlgebra alpha] [KSSeparation alpha]
3   [KSSeparationStrict alpha] :
4   ∃ Theta : alpha → R,
5     (∀ a b : alpha, a ≤ b ↔ Theta a ≤ Theta b) ∧
6     Theta ident = 0 ∧
7     ∀ x y : alpha, Theta (op x y) = Theta x + Theta y := by
8   -- Use Theta_cuts (the Dedekind-cuts construction)
9   obtain ⟨a0, ha0⟩ := <witness for non-trivial element>
10  refine <Theta_cuts a0 ha0, order_preservation, identity, additivity>
```

Remark 4.11 (Comparison to Grid Proof). The cuts proof is significantly more compact:

- **Grid proof:** ~2000+ lines (induction machinery, extension lemmas, path independence)
- **Cuts proof:** ~500 lines (direct construction, no induction)

However, the grid proof more closely follows K&S's original argument structure (A/B/C partition, δ -choice), while the cuts proof uses the standard Hölder technique from ordered group theory.

Corollary 4.12 (op.comm_of_associativity). *Lines 87–92, Additive/Proofs/GridInduction/-Main.lean*

Commutativity follows from the representation theorem.

Listing 15: Commutativity from representation

```

1 theorem op_comm_of_associativity
2   (alpha : Type*) [KnuthSkillingMonoidBase alpha] [KSSeparation alpha]
3   [RepresentationGlobalization alpha] :
4   ∀ x y : alpha, op x y = op y x := by
5   classical
6   obtain <Theta, hTheta_order, _, hTheta_add> := associativity_representation (
7     alpha := alpha)
    exact commutativity_from_representation Theta hTheta_order hTheta_add

```

5 The Product Theorem (Appendix B)

Files:

- Multiplicative/Main.lean (K&S's Appendix B pipeline via Appendix A)
- Multiplicative/Proofs/Direct/DirectProof.lean (Alternative: direct algebraic path)
- Multiplicative/ScaledMultRep.lean (Common interface for both paths)

5.1 Two Complete Proof Paths

Like Appendix A, the formalization provides **two independent proofs** of Appendix B's conclusion. Both paths arrive at the same result: the tensor operation \otimes on positive reals equals multiplication up to a global scale constant.

5.1.1 Path 1: K&S's Actual Derivation (Recommended)

Multiplicative/Main.lean follows K&S's paper exactly: “apply Appendix A again to \otimes ”. This path uses AdditiveOrderIsoRep (from Appendix A) to derive the product equation, then solves it to show \otimes is scaled multiplication.

5.1.2 Path 2: Alternative (Direct Algebraic Proof)

Multiplicative/Proofs/Direct/DirectProof.lean provides a direct algebraic proof that any tensor satisfying distributivity (Axiom 3) and associativity (Axiom 4) must be scaled multiplication.

Remark 5.1 (Why Two Paths?). • **Path 1** assumes existence of AdditiveOrderIsoRep for the tensor (“apply Appendix A again”)

- **Path 2** derives the same result directly from distributivity + associativity axioms
- Both arrive at the same conclusion: \otimes is scaled multiplication
- **Note:** This is NOT “Aczél’s derivation of probability theory” (a separate classical approach); it’s just an alternative proof technique for K&S’s Appendix B

5.2 Common Interface: ScaledMultRep

Both paths provide the `ScaledMultRep` interface, which captures the OUTPUT of Appendix B:

Listing 16: ScaledMultRep interface (Multiplicative/ScaledMultRep.lean:44)

```

1 structure ScaledMultRep (tensor : PosR → PosR → PosR) where
2   C : R           -- The scale constant C > 0
3   C_pos : 0 < C
4   tensor_eq : ∀ x y : PosR,
5     ((tensor x y) : R) = ((x : R) * (y : R)) / C

```

Design principle: Like `AdditiveOrderIsoRep` for Appendix A, this interface captures WHAT Appendix B proves without depending on HOW it was proven. Downstream code should depend on `ScaledMultRep`, not on specific proof paths.

Constructors:

- `scaledMultRep_of_additiveOrderIsoRep`: K&S path (uses Appendix A)
- `scaledMultRep_of_tensorRegularity`: Direct path (bypasses Appendix A)
- `scaledMultRep_of_assoc_distrib_comm`: Minimal assumptions (assoc + distrib + comm)

5.3 Product-Side Symmetries (3–4)

K&S paper location: Symmetry 3 appears at equation (7) on page 6 (arxiv.tex lines 462–467), Axiom 3 at equation (24) on page 9 (arxiv.tex lines 566–572).

Before applying Appendix A, K&S work with lattice elements and the direct-product operator \times . **Symmetry 3** states that \times is (right-)distributive over the join \sqcup :

$$(x \times t) \sqcup (y \times t) = (x \sqcup y) \times t$$

After Appendix A provides the representation $\Theta : \alpha \rightarrow \mathbb{R}$, we work with graded measures and the tensor operation \otimes . **Axiom 3** is the graded version:

$$(x \otimes t) \oplus (y \otimes t) = (x \oplus y) \otimes t$$

After moving to real numbers via Θ , this becomes (Appendix B, arxiv.tex line 661):

$$x \otimes t + y \otimes t = (x + y) \otimes t$$

where $+$ is real addition (since \oplus has been identified with $+$ by Appendix A).

Symmetry 4 (Product Associativity): $(u \otimes v) \otimes w = u \otimes (v \otimes w)$

Formalization note: We have lattice-level Symmetry 3 in `DirectProduct.prod_sup_left`:

```

1 -- Multiplicative/DirectProduct.lean, line 42
2 prod_sup_left : ∀ a1 a2 : alpha, ∀ b : beta,
3   prod (a1 || a2) b = prod a1 b || prod a2 b -- // denotes sup

```

At the graded level, we define `DistributesOverAdd` as a property:

- `DistributesOverAdd` is a predicate that a tensor may or may not satisfy
- `TensorAlgebra` is the bundled class including this property
- The derivation of `DistributesOverAdd` from `prod_sup_left` is now **fully formalized** in `DistributivityDerivation.lean`

Listing 17: Distributivity property (Multiplicative/Basic.lean:68)

```

1  -- Defines the PROPERTY (not an axiom, just a predicate)
2  def DistributesOverAdd (tensor : PosR → PosR → PosR) : Prop :=
3    ∀ x y t : PosR, tensor (addPos x y) t = addPos (tensor x t) (tensor y t)
4
5  -- Then we ASSUME some tensor satisfies this property:
6  variable (hDistrib : DistributesOverAdd tensor)

```

Remark 5.2 (Connection between lattice and graded levels). In K&S's development:

1. Symmetry 3 is stated at the lattice level: $(x \times t) \sqcup (y \times t) = (x \sqcup y) \times t$
2. After Appendix A provides the representation Θ , this gives distributivity at the graded level
3. The graded tensor satisfies `DistributesOverAdd`

Current state: We have *all three* levels formalized:

- Lattice level: `DirectProduct.prod_sup_left` (Multiplicative/DirectProduct.lean)
- Graded level: `DistributesOverAdd` (Multiplicative/Basic.lean)
- Bridge: `distributes_over_add_from_lattice` (Multiplicative/DistributivityDerivation.lean)

Derivation (COMPLETE): The theorem `distributes_over_add_from_lattice` proves that `DistributesOverAdd` follows from:

1. `prod_sup_left`: Lattice-level distributivity
2. `disjoint_prod_left`: Disjointness preservation
3. `sum_rule`: Valuation additivity on disjoint events (from `CoxConsistency`)
4. `RectTensorCompatible`: Bridge predicate $v(\text{prod } a b) = \text{tensor}(v(a), v(b))$

This shows that scalar distributivity is **derived**, not assumed!

Remark 5.3 (Unbundled Tensor Predicates and `TensorAlgebra`). **Lines 59–123, Multiplicative/Basic.lean**

Following the unified axiom organization, tensor properties have both unbundled predicates and a bundled class:

Unbundled predicates:

- `TensorAssoc tensor` (line 72): Associativity of \otimes
- `TensorPos tensor` (line 77): Positivity-preserving

- `TensorStrictMonoLeft tensor` (line 81): Strict monotonicity in left argument
- `TensorStrictMonoRight tensor` (line 85): Strict monotonicity in right argument
- `DistributesOverAdd tensor` (line 68): Distributivity over $+$

Bundled class (line 103):

```

1 class TensorAlgebra (tensor : PosR → PosR → PosR) : Prop where
2   distributes : DistributesOverAdd tensor
3   assoc : TensorAssoc tensor
4   pos : TensorPos tensor

```

Convenience theorem (line 247): `productEquation_of_tensorAlgebra` provides an ergonomic entry point for proofs that use all the bundled axioms together.

Design principle: Use unbundled predicates (e.g., `hDistrib : DistributesOverAdd tensor`) when tracking minimal hypotheses. Use `[TensorAlgebra tensor]` for ergonomic access in longer proofs.

5.4 Product Equation

Appendix B shows \otimes must be multiplication up to a global scale.

Theorem 5.4 (Psi_is_exp). *Line 43, Multiplicative/Main.lean*

The inverse representation $\Psi = \Theta^{-1}$ is exponential: $\Psi(x) = C \cdot e^{Ax}$ for some constants $C > 0$ and A .

Listing 18: Appendix B: Ψ is exponential

```

1 theorem Psi_is_exp
2   (hRep : AdditiveOrderIsoRep PosR tensor)
3   (hDistrib : DistributesOverAdd tensor) :
4   ∃ (C A : R), 0 < C ∧ ∀ x : R, Derived.Psi hRep x = C * R.exp (A * x)
5   := by
6   refine
7     productEquation_solution_of_continuous_strictMono
8     (hEq := productEquation_Psi (tensor := tensor) hRep hDistrib)
9     (hPos := fun x => Derived.Psi_pos (tensor := tensor) hRep x)
10    (hCont := Derived.Psi_continuous (tensor := tensor) hRep)
11    (hMono := Derived.Psi_strictMono (tensor := tensor) hRep)

```

Remark 5.5 (The Functional Equation Proof). The proof delegates to:

`productEquation_solution_of_continuous_strictMono`
(line 294, Multiplicative/FunctionalEquation.lean)

This proves a **classical result from functional equations theory**:

Statement: If $\Psi : \mathbb{R} \rightarrow \mathbb{R}$ satisfies the product equation

$$\Psi(\tau + \xi) + \Psi(\tau + \eta) = \Psi(\tau + \zeta(\xi, \eta))$$

for all $\tau, \xi, \eta \in \mathbb{R}$, and if Ψ is positive, continuous, and strictly monotone, then $\Psi(x) = C \cdot e^{Ax}$ for some constants $C > 0$ and A .

Key steps (561 lines):

1. Extract shift constant: $a := \zeta(0, 0)$ gives $\Psi(x + a) = 2\Psi(x)$

2. Extend to powers: $\Psi(x + na) = 2^n \Psi(x)$ for all $n \in \mathbb{Z}$
3. Extend to rationals: $\Psi(x + (m/n)a) = 2^{m/n} \Psi(x)$ for all $m/n \in \mathbb{Q}$
4. Use continuity + density: Extend to all reals
5. Conclude: $\Psi(x) = C \cdot 2^{x/a} = C \cdot e^{(\ln 2/a) \cdot x}$

The continuity and monotonicity hypotheses are **derived** (not assumed) from the order isomorphism $\Theta : \text{PosReal} \simeq_o \mathbb{R}$ established in Appendix A:

- **Psi.strictMono** (line 148, *Multiplicative/Basic.lean*): Since $\Psi := \Theta^{-1}$ and Θ is an order isomorphism, Θ^{-1} is strictly monotone.
- **Psi_continuous** (line 154, *Multiplicative/Basic.lean*): Order isomorphisms $\mathbb{R} \simeq_o \mathbb{R}$ are continuous (order topology).

Why Exponential Ψ Implies Tensor = Scaled Multiplication

Given: $\Theta(x \otimes y) = \Theta(x) + \Theta(y)$ (additivity) and $\Psi = \Theta^{-1}$ with $\Psi(z) = C \cdot e^{Az}$

Derivation:

$$\begin{aligned} x &= \Psi(\Theta(x)) = C \cdot e^{A \cdot \Theta(x)} \Rightarrow e^{A \cdot \Theta(x)} = x/C \\ x \otimes y &= \Psi(\Theta(x \otimes y)) = \Psi(\Theta(x) + \Theta(y)) \\ &= C \cdot e^{A(\Theta(x) + \Theta(y))} = C \cdot e^{A \cdot \Theta(x)} \cdot e^{A \cdot \Theta(y)} \\ &= C \cdot (x/C) \cdot (y/C) = \frac{x \cdot y}{C} \end{aligned}$$

Conclusion: $x \otimes y = (x \cdot y)/C$ (Lean: `tensor_coe_eq_mul_div_const`, line 61)

Theorem 5.6 (`tensor_mul_rule_normalized`). Line 105, *Multiplicative/Main.lean*

The tensor operation is multiplication up to a global constant: $(x \otimes y)/C = (x/C) \cdot (y/C)$.

Listing 19: Product rule (normalized)

```

1 theorem tensor_mul_rule_normalized
2   (hRep : AdditiveOrderIsoRep PosR tensor)
3   (hDistrib : DistributesOverAdd tensor) :
4   ∃ C : R, 0 < C /\ 
5     (∀ x : PosR, 0 < ((x : R) / C)) /\ 
6     (∀ x y : PosR,
7      ((tensor x y : PosR) : R) / C = (((x : R) / C) * ((y : R) / C)))

```

6 The Variational Theorem (Appendix C)

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Variational/Main.lean

6.1 Two Logical Steps: Path A and Path B

Appendix C separates cleanly into two steps.

Path A (rigidity / solution): assume the functional equation holds globally and add an explicit regularity gate (we use measurability of H'). Then the only solutions are logarithmic (see `variationalEquation_solution_measurable`, lines 315–380 in `Variational/Main.lean`).

Path B (derivation / motivation): compute the coordinate derivative of a separable product Lagrangian at a stationary point and obtain the separated equation that motivates the functional equation (see `section PathB` starting at line 625 and `lagrange_coordinate_deriv_eq` at line 699 in `Variational/Main.lean`). Upgrading “holds at a stationary point” to a global equation is an explicit universality/richness premise (Gate G).

6.2 Variational Functional Equation

What this is about: K&S derive the entropy form $H(m) = A + Bm + C(m \log m - m)$ (the negative of Shannon/Boltzmann entropy) from a variational principle. The key functional equation comes from maximizing H subject to constraints.

The equation: The derivative $H'(m)$ must satisfy:

$$H'(m_x \cdot m_y) = \lambda(m_x) + \mu(m_y)$$

where m_x, m_y are probability masses (positive reals).

Intuition: This says the potential function separates multiplicatively - the derivative at a product decomposes into separate contributions from each factor.

Definition 6.1 (VariationalEquation). Lines 201–202, Variational/Main.lean

Listing 20: Variational equation definition

```
1 def VariationalEquation (H' : R → R) : Prop :=
2   ∀ m_x m_y : R, 0 < m_x → 0 < m_y → H' (m_x * m_y) = lam m_x + mu m_y
```

6.3 Main Theorem

Result: The only measurable solutions to the variational equation are logarithmic.

Why this matters: This shows that the entropy form is **uniquely determined** (up to constants) by the variational principle plus measurability. You can't have some other weird function satisfy the constraints.

Proof strategy: Transform the multiplicative equation $H'(m_x \cdot m_y) = \lambda(m_x) + \mu(m_y)$ into Cauchy's additive equation $f(u + v) = f(u) + f(v)$ by setting $u = \log m$. Measurable solutions to Cauchy's equation are linear [6], giving $H'(m) = B + C \log m$.

Theorem 6.2 (variationalEquation_solution_measurable). Lines 310–375, Variational/-Main.lean

If H' satisfies the variational equation and is Borel-measurable, then:

$$H'(m) = B + C \cdot \log(m)$$

for some constants B, C .

Listing 21: Appendix C main theorem

```

1 theorem variationalEquation_solution_measurable
2   (H' : R → R) (lam mu : R → R)
3   (hMeas : Measurable H')
4   (hV : VariationalEquation H' lam mu) :
5   ∃ B C : R, ∀ m : R, 0 < m → H' m = B + C * R.log m := by
6   -- Step 1: Extract the common core phi from lam and mu
7   obtain <phi, c1, c2, hphi1, hlam, hmu> := hV.∃_common_core
8   -- Step 2-7: Transform to Cauchy equation and apply linear solution
9   ...

```

6.4 The Entropy Form

What this is: The classical Shannon/Boltzmann entropy appears as the antiderivative of the logarithmic solution.

Derivation: Integrating $H'(m) = B + C \log(m)$ with respect to m :

$$H(m) = \int (B + C \log m) dm = A + Bm + C(m \log m - m)$$

where the integration constant is A .

Physical interpretation: The term $m \log m$ is (up to sign and constants) the Shannon entropy $-\sum p_i \log p_i$ for discrete distributions, or the Boltzmann entropy $-\int p(x) \log p(x) dx$ for continuous distributions. The other terms (A, Bm) are normalization and constraint adjustments.

Why it matters: This shows that entropy **isn't** an axiom** - it's derived from the variational principle applied to the K&S probability framework.

Definition 6.3 (entropyForm). Line 485, Variational/Main.lean

Listing 22: Entropy form

```

1 noncomputable def entropyForm (A B C : R) : R → R :=
2   fun m => A + B * m + C * (m * R.log m - m)

```

Definition 6.4 (entropyDerivative). Line 287, Variational/Main.lean

The expected derivative of the entropy form.

Listing 23: Expected derivative: $B + C \log m$

```

1 noncomputable def entropyDerivative (B C : R) : R → R :=
2   fun m => B + C * R.log m

```

Theorem 6.5 (entropyForm_deriv). Lines 488–512, Variational/Main.lean

The entropy form has derivative $H'(m) = B + C \log m$.

Listing 24: Proof that entropy form has the correct derivative

```

1 theorem entropyForm_deriv (A B C : R) {m : R} (hm : 0 < m) :
2   HasDerivAt (entropyForm A B C) (entropyDerivative B C m) m := by
3   unfold entropyForm entropyDerivative
4   -- d/dm [A + Bm + C(m log m - m)] = B + C(log m + 1 - 1) = B + C log m
5   ...

```

Remark 6.6 (What this proves). The theorem `entropyForm_deriv` proves that:

$$\frac{d}{dm} [A + Bm + C(m \log m - m)] = B + C \log m$$

This verifies that integrating the logarithmic solution $H'(m) = B + C \log m$ (from the variational equation) gives the entropy form $H(m) = A + Bm + C(m \log m - m)$.

7 Divergence (K&S Section 6)

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Information/Divergence.lean

What this is: The divergence $\phi(w, u)$ measures the “distance” between two measure assignments w and u . It quantifies the information-theoretic cost of using measure u when the “true” measure is w .

Connection to Appendix C (Entropy Form): The divergence is a **special case** of the entropy form from the variational theorem:

$$\begin{aligned} H(m) &= A + Bm + C(m \log m - m) \quad (\text{Appendix C}) \\ \phi(w, u) &= u - w + w \log(w/u) \quad (\text{Divergence}) \end{aligned}$$

Setting $A = u$, $B = -\log(u)$, $C = 1$ in the entropy form gives the divergence (K&S Eq. 44). The critical point analysis from Appendix C proves that $\phi(w, u)$ is minimized when $w = u$.

Note: This is **atom divergence** for general real-valued measures. The specialization to **probability distributions** happens in Section 11, after Section 9 derives what probability distributions are.

Key properties:

- **Non-negative:** $\phi(w, u) \geq 0$, with equality iff $w = u$ (formalized in `atomDivergence_nonneg`, lines 102–120)
- **Asymmetric:** $\phi(w, u) \neq \phi(u, w)$ in general (it’s NOT a distance metric)
- **Connects variational calculus to information theory:** Bridge between Appendix C and Section 8

Forward reference: This atom divergence will be specialized to **probability distributions** in Section 11 (Information and Entropy), giving the Kullback-Leibler divergence formula.

Definition 7.1 (atomDivergence). Lines 68–69, Divergence.lean

The per-atom divergence: $\phi(w, u) = u - w + w \log(w/u)$.

Listing 25: Atom divergence

```
1 noncomputable def atomDivergence (w u : R) : R :=
2   u - w + w * log (w / u)
```

Theorem 7.2 (atomDivergence_nonneg). Lines 102–120, Divergence.lean

Listing 26: Divergence non-negativity

```

1 theorem atomDivergence_nonneg (w u : R) (hw : 0 < w) (hu : 0 < u) :
2   0 ≤ atomDivergence w u := by
3   unfold atomDivergence
4   -- Rewrite as  $w * (u/w - 1 - \log(u/w))$  and use log inequality
5   let s := u / w
6   have hs : 0 < s := div_pos hu hw
7   have hrewrite : u - w + w * log (w / u) = w * (s - 1 - log s) := by ...
8   rw [hrewrite]
9   exact mul_nonneg (le_of_lt hw) (log_ineq s hs)

```

Theorem 7.3 (`atomDivergence_eq_zero_iff`). *Lines 123–159, Divergence.lean*

Divergence equals zero if and only if $w = u$.

Listing 27: Divergence equals zero iff $w = u$ (Divergence.lean:123)

```

1 theorem atomDivergence_eq_zero_iff (w u : R) (hw : 0 < w) (hu : 0 < u) :
2   atomDivergence w u = 0 ↔ w = u := by
3   constructor
4   . -- If  $\phi(w, u) = 0$ , then  $w = u$ 
5   intro h
6   let s := u / w
7   have hs : 0 < s := div_pos hu hw
8   --  $\phi(w, u) = w * (s - 1 - \log s) = 0$ 
9   --  $w > 0$  implies  $s - 1 - \log s = 0$ 
10  -- But  $s - 1 - \log s > 0$  for  $s \neq 1$  (strict log inequality)
11  -- So  $s = 1$ , hence  $u = w$ 
12  ...
13  .
14  intro heq
15  rw [heq]
16  exact atomDivergence_self u hu

```

8 Conditional Probability (K&S Section 7)

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Probability/ConditionalProbability/Basic.lean

What this section does: K&S Section 7 derives **probability calculus** from first principles. This is the crucial step that takes us from general measures (Sections 1–6) to **probability distributions** (normalized measures).

Starting with conditional plausibility as a **bivaluation** $p(x|t)$ (a function taking pairs of lattice elements to reals), K&S introduces **Axiom 5 (Chaining Associativity)** and proves:

1. The **chain-product rule**: $\Pr(a|c) = \Pr(a|b) \cdot \Pr(b|c)$ for chains $a \leq b \leq c$
2. **Bayes' theorem**: $\Pr(x|\theta) \cdot \Pr(\theta) = \Pr(\theta|x) \cdot \Pr(x)$
3. **Probability as a ratio**: $\Pr(x|t) = \frac{m(x \wedge t)}{m(t)}$ (K&S Eq. 53)

Key insight: The SAME functional equation from Appendix B (product equation) reappears here! Axiom 5 + sum rule forces the chaining operation to be multiplication (up to scale).

Deliverable: This section establishes that **probability is a normalized measure** – “simply the shape of the confined measure, automatically normalized to unit mass” (K&S, Section 7.3). This gives us **probability distributions**, which are used in Section 11.

8.1 Structural Change: From Linear Order to Lattice

Remark 8.1 (Different Type Structure). K&S Section 7 operates on a **different type** than Sections 1–6:

- **Sections 1–6** (K&S algebra): `[LinearOrder α]` - measures on linearly ordered values
- **Section 7** (Bivaluation): `[Lattice α]` `[BoundedOrder α]` - probability on lattice of events

This reflects K&S's conceptual shift: earlier sections study **measure values** (which are linearly ordered reals), while Section 7 studies **conditional probability on events** (which form a lattice).

The Lean formalization respects the **logical dependency order**, not K&S's presentation order. Section 7 requires lattice operations (\wedge , \vee , \perp , \top) that weren't needed in Sections 1–6.

Lattice hierarchy in the code:

- Bivaluation structure (line 59): `[Lattice α]` - general lattice
- Main theorems (chain-product, Bayes): `[DistribLattice α]` - needs distributivity
- Optional theorems (`sumRule_general`, `complementRule`): `[BooleanAlgebra α]` - needs complements

Mathematical generalization: K&S works with propositions (Boolean), but the Lean formalization proves the core probability calculus works on **any distributive lattice**. Boolean structure is only needed for complement operations.

8.2 Bivaluation and Axiom 5

Definition 8.2 (Bivaluation). **Lines 59–73, Basic.lean**

A bivaluation $p : \alpha \rightarrow \alpha \rightarrow \mathbb{R}$ represents conditional plausibility on a lattice with:

- **Positivity:** $p(x|t) > 0$ when $\perp < x \leq t$
- **Sum rule:** $p(x \vee y|t) = p(x|t) + p(y|t)$ for disjoint x, y
- **Context intersection:** $p(x|t) = p(x \wedge t|t)$ (implicit in K&S)

Remark 8.3 (Lattice Structure on Events, Not Context). Note that the sum rule applies to the **first argument** (the event), not the context:

$$p(x \vee y | t) = p(x | t) + p(y | t)$$

The context t stays **fixed** while events are decomposed via the lattice join \vee . This matches the standard probability identity $P(A \cup B | C) = P(A | C) + P(B | C)$ for disjoint A, B . The lattice operations (\vee, \wedge, \perp) describe the *event algebra*; the context is just a parameter.

Definition 8.4 (Axiom 5: Chaining Associativity). **Lines 109–128 (ChainingOp structure), Basic.lean**

The chaining operation $\odot : \mathbb{R} \rightarrow \mathbb{R} \rightarrow \mathbb{R}$ on plausibility values satisfies:

Listing 28: ChainingOp structure (Basic.lean:124)

```

1 structure ChainingOp where
2   chain : R → R → R
3   chain_assoc : ∀ x y z, chain (chain x y) z = chain x (chain y z)
4   chain_strictMono_left : ∀ z, 0 < z → StrictMono (fun x => chain x z)
5   chain_strictMono_right : ∀ x, 0 < x → StrictMono (fun z => chain x z)
6   chain_pos : ∀ x y, 0 < x → 0 < y → 0 < chain x y
7   chain_distrib_left : ∀ a b t, 0 < a → 0 < b → 0 < t →
8     chain a t + chain b t = chain (a + b) t

```

Formulation: For a chain $a < b < c < d$:

$$(p(a|b) \odot p(b|c)) \odot p(c|d) = p(a|b) \odot (p(b|c) \odot p(c|d))$$

Definition 8.5 (Chain Rule). Line 219, Basic.lean

The chain rule connects the chaining operation to the bivaluation:

Listing 29: ChainingAssociativity class (Basic.lean:219)

```

1 class ChainingAssociativity (alpha : Type*) [Lattice alpha] [BoundedOrder alpha]
2   (B : Bivaluation alpha) where
3   chainOp : ChainingOp
4   chain_rule : ∀ a b c : alpha, a ≤ b → b ≤ c → Bot < a →
5     B.p a c = chainOp.chain (B.p a b) (B.p b c)

```

This says: $p(a|c) = p(a|b) \odot p(b|c)$ for chains $a \leq b \leq c$.¹

8.3 The Product Equation Reappears

Lines 245–314, Basic.lean

K&S's brilliant observation: combining chaining associativity with the sum rule gives the **exact same product equation** as Appendix B!

Proof sketch:

1. Let Θ be the function such that $\Theta(p(a|b) \odot p(b|c)) = \Theta(p(a|b)) + \Theta(p(b|c))$
2. Define $\Psi = \Theta^{-1}$
3. The sum rule says: for disjoint x, y with intermediate context,

$$\text{chain}(a, t) + \text{chain}(b, t) = \text{chain}(a + b, t)$$

This is **left-distributivity over addition** - exactly the Appendix B hypothesis!

4. Therefore: $\Psi(\tau + \xi) + \Psi(\tau + \eta) = \Psi(\tau + \zeta(\xi, \eta))$ where $\zeta(\xi, \eta) = \Theta(\Psi(\xi) + \Psi(\eta))$
5. By Appendix B: $\Theta = A \cdot \log$ for some $A > 0$
6. Hence the chaining operation is: $\text{chain}(x, y) = \frac{x \cdot y}{K}$ for some $K > 0$

¹K&S uses “interval notation” $[x, y]$ throughout Section 7 without formally defining intervals as mathematical objects. They write $\alpha = [x, y]$, $\beta = [y, z]$, etc., and speak of “concatenating intervals” $[x, y] \circ [y, z] = [x, z]$. The chaining operation \odot then acts on the plausibility *values* of these intervals: $p(\alpha) \odot p(\beta)$. Our formalization sidesteps this implicit interval semantics by working directly with lattice elements: the “interval $[a, b]$ ” is represented implicitly by the pair (a, b) with constraint $a \leq b$. The chain rule then states $p(a|c) = p(a|b) \odot p(b|c)$ for $a \leq b \leq c$, which captures the compositional structure without reifying intervals as first-class objects.

8.4 Chain-Product Rule

Theorem 8.6 (chainProductRule). *Line 345, Basic.lean*

For chains $a \leq b \leq c$ in a lattice with normalized bivaluation ($p(t|t) = 1$):

$$\Pr(a|c) = \Pr(a|b) \cdot \Pr(b|c)$$

Listing 30: chainProductRule (Basic.lean:345)

```

1 theorem chainProductRule
2   {alpha : Type*} [DistribLattice alpha] [BoundedOrder alpha]
3   (B : Bivaluation alpha) [CA : ChainingAssociativity alpha B]
4   (hNormalized : ∀ t : alpha, Bot < t → B.p t t = 1) :
5   ∀ a b c : alpha, a ≤ b → b ≤ c → Bot < a →
6     B.p a c = B.p a b * B.p b c

```

Proof strategy:

- Appendix B gives: $\text{chain}(x, y) = (x \cdot y)/K$ for some $K > 0$
- Normalization at (a, a, a) forces $K = 1$: since $p(a|a) = 1$, we have $1 = \text{chain}(1, 1) = 1/K$
- Therefore: $p(a|c) = \text{chain}(p(a|b), p(b|c)) = p(a|b) \cdot p(b|c)$

8.5 Bayes' Theorem

Theorem 8.7 (bayesTheorem). *Line 421, Basic.lean*

For $x, \theta \leq t$ in a distributive lattice:

$$\Pr(x|\theta) \cdot \Pr(\theta|t) = \Pr(\theta|x) \cdot \Pr(x|t)$$

Listing 31: bayesTheorem (Basic.lean:421)

```

1 theorem bayesTheorem
2   {alpha : Type*} [DistribLattice alpha] [BoundedOrder alpha]
3   (B : Bivaluation alpha) [CA : ChainingAssociativity alpha B]
4   (hNormalized : ∀ t : alpha, Bot < t → B.p t t = 1)
5   (x theta t : alpha) (hxtheta_pos : Bot < x □ theta) (hx : x ≤ t) (htheta :
6     theta ≤ t)
7   (hx_pos : Bot < x) (htheta_pos : Bot < theta) :
      B.p x theta * B.p theta t = B.p theta x * B.p x t

```

Proof: Both sides equal $\Pr(x \wedge \theta|t)$ by the product rule and commutativity of \wedge .

8.6 Probability as Ratio of Measures

Theorem 8.8 (prob_eq_measure_ratio). *Line 462, Basic.lean*

Define the **unconditional measure** by $m(x) := p(x|\top)$. Then for any context $t \neq \perp$:

$$\Pr(x|t) = \frac{m(x \wedge t)}{m(t)}$$

Listing 32: prob_eq_measure_ratio (Basic.lean:462)

```

1 theorem prob_eq_measure_ratio
2   {alpha : Type*} [DistribLattice alpha] [BoundedOrder alpha]
3   (B : Bivaluation alpha) [CA : ChainingAssociativity alpha B]
4   (hNormalized : ∀ t : alpha, Bot < t → B.p t t = 1) :
5   ∀ x t : alpha, t ≠ Bot → B.p x t = baseMeasure B (x □ t) / baseMeasure B t

```

This single formula subsumes the sum rule, chain-product rule, and range $[0, 1]$. Probability is simply the **ratio of measures** — “the elementary calculus of proportions of measure” (K&S, Section 7.3).

8.7 baseMeasure Satisfies Measure Axioms

Lines 559–576, Basic.lean

The derived `baseMeasure` satisfies the classical measure axioms:

Theorem 8.9 (`baseMeasure_satisfies_measure_axioms`). *For a normalized Bivaluation ($p(t|t) = 1$ for $t > \perp$), `baseMeasure` is a probability measure:*

1. $m(\perp) = 0$ (empty set has measure zero)
2. Finite additivity: $m(x ∨ y) = m(x) + m(y)$ for disjoint x, y
3. Non-negativity: $0 \leq m(x)$
4. Normalization: $m(\top) = 1$

Listing 33: baseMeasure_satisfies_measure_axioms (Basic.lean:559)

```

1 theorem baseMeasure_satisfies_measure_axioms
2   (hNormalized : ∀ t : alpha, Bot < t → B.p t t = 1)
3   (hTop : (Top : alpha) ≠ Bot) :
4   baseMeasure B Bot = 0 /\
5   (∀ x y : alpha, Disjoint x y →
6     baseMeasure B (x ∪ y) = baseMeasure B x + baseMeasure B y) /\
7   (∀ x : alpha, 0 ≤ baseMeasure B x) /\
8   baseMeasure B Top = 1

```

Key point: For finite Boolean algebras, finite additivity is equivalent to σ -additivity, so this is a bona fide probability measure in the Kolmogorov sense [8].

Remark 8.10 (Additional Measure Properties). The formalization also proves:

- **Inclusion-exclusion** (`baseMeasure_inclusion_exclusion`, line 588):

$$m(x ∨ y) + m(x ∧ y) = m(x) + m(y)$$

- **Complement rule** (`baseMeasure_compl_normalized`, line 640):

$$m(x^c) = 1 - m(x)$$

- **Subadditivity** (`baseMeasure_subadditive`, line 649):

$$m(x ∨ y) \leq m(x) + m(y)$$

- **ENNReal version** (`baseMeasureENNReal`, line 673): For Mathlib compatibility

9 σ -Completeness and σ -Additivity (Extension)

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Core/ScaleCompleteness.lean

Purpose: K&S develop a *finite* additivity theory on event algebras. To connect this to Kolmogorov-style probability measures on σ -algebras, we need a **countable** closure story:

- events must support countable joins (“ σ -completeness”),
- the valuation must preserve limits of increasing chains (Scott continuity), and
- the scale (the image of Θ) must be sequentially complete so that sup values exist in the scale even if they need not exist inside the original algebra.

This file packages these as **minimal, explicit** additional axioms and proves the corresponding σ -additivity theorem for disjoint unions.

9.1 σ -Complete Events

Definition 9.1 (SigmaCompleteEvents). Lines 144–152, Core/ScaleCompleteness.lean

Extends PlausibilitySpace with a countable supremum operator `iSup : (Nat → E) → E` satisfying the usual upper-bound / least-upper-bound laws.

9.2 Sequential Completeness of the Scale

Definition 9.2 (KSScaleComplete). Lines 81–89, Core/ScaleCompleteness.lean

Given a representation `R : RepresentationResult S` (Appendix A output), `KSScaleComplete S R` states that $\Theta(S) \subseteq \mathbb{R}$ is closed under suprema of bounded increasing ω -sequences.

9.3 Scott Continuity (Countable Directed Limits)

Definition 9.3 (KSScottContinuous). Lines 203–211, Core/ScaleCompleteness.lean

Scott continuity (phrased via $R.\Theta$) states that for a monotone sequence of events $f : \mathbb{N} \rightarrow E$,

$$\Theta(v(\sup_n f)) = \sup_n \Theta(v(f_n)).$$

9.4 Main Theorem: K&S σ -Additivity

Theorem 9.4 (ks_sigma_additive). Lines 360–418, Core/ScaleCompleteness.lean

Under:

- `SigmaCompleteEvents E`
- a normalized representation `R : NormalizedRepresentationResult S`
- `KSScaleComplete S R.toRepresentationResult`
- `KSScottContinuous E S R.toRepresentationResult`
- pairwise disjointness of $f : Nat \rightarrow E$

the induced real-valued measure $\mu := \Theta \circ v$ is countably additive:

$$\mu\left(\sup_n f_n\right) = \sum_{n=0}^{\infty} \mu(f_n).$$

Remark 9.5 (Mathlib Bridge). For a bridge to mathlib's Measure / ProbabilityMeasure APIs, see: Mettapedia/ProbabilityTheory/KnuthSkilling/Bridges/MathlibProbability.lean.

10 Information and Entropy (K&S Section 8)

File: Mettapedia/ProbabilityTheory/KnuthSkilling/Information/InformationEntropy.lean

What this section does: K&S Section 8 takes **special cases** of the variational potential H from Appendix C, specialized to probability distributions (normalized measures from Section 7).

Key point: Shannon entropy is **derived**, not just defined. It emerges as an “inevitable consequence of seeking a variational quantity” (K&S, Section 8.2).

Unified view (project-wide, optional): The repo also includes axiomatic entropy (Faddeev / Shannon–Khinchin) and bridges to mathlib's measure-theoretic KL; the curated entrypoint Mettapedia/InformationTheory/EntropyKL.lean imports the unified finite-discrete story and records the main equivalence theorems (see Appendix: Supplementary Derivation Paths).

10.1 From Atom Divergence to KL Divergence

The key step: Now that we have probability distributions from Section 9, we can specialize the atom divergence from Section 8 to normalized measures.

For probability distributions $P = (p_1, \dots, p_n)$ and $Q = (q_1, \dots, q_n)$ where $\sum p_i = 1$ and $\sum q_i = 1$:

$$\begin{aligned} \sum_i \phi(p_i, q_i) &= \sum_i (q_i - p_i + p_i \log(p_i/q_i)) \\ &= \underbrace{\sum_i q_i}_{=1} - \underbrace{\sum_i p_i}_{=1} + \sum_i p_i \log(p_i/q_i) \\ &= \sum_i p_i \log(p_i/q_i) = D_{KL}(P\|Q) \end{aligned}$$

This is the **Kullback-Leibler divergence** (K&S Eq. 54).

Formalized in Lean (`klDivergence_from_divergence_formula`, lines 324–338, `Information/InformationEntropy.lean`):

```

1 theorem klDivergence_from_divergence_formula (P Q : ProbDist n)
2   (hQ_pos : ∀ i, P.p i ≠ 0 → 0 < Q.p i) :
3     klDivergence P Q hQ_pos =
4       sum i, atomDivergence (P.p i) (Q.p i) - (sum i, Q.p i - sum i, P.p i)

```

The proof uses the normalization constraints: $\sum(q_i - p_i) = 1 - 1 = 0$, so the linear terms cancel.

10.2 Derivation Chain

From Appendix C to Shannon Entropy:

1. **Appendix C** establishes the general variational form for any measure:

$$H(m) = A + B \cdot m + C \cdot (m \log m - m)$$

2. **Section 8** specializes to atom divergence: $\phi(w, u) = u - w + w \log(w/u)$

3. **Section 9** proves that probability is a normalized measure: $\Pr(x|t) = m(x \wedge t)/m(t)$
4. **Section 11 (this section):** For probability distributions, divergence simplifies to KL divergence
5. **Section 8.2 (Entropy):** To quantify uncertainty, we require:
 - Zero uncertainty when one $p_k = 1$ (fully determined state)
 - This forces: $A_k = 0$ and $B_k = C$
 - Setting $C = -1$ (conventional scale) gives:

$$S(p) = - \sum_k p_k \log p_k$$

This is **Shannon entropy** - not assumed, but **derived from the variational principle**.

10.3 Shannon's Three Properties

K&S claim these properties are “inevitable consequences” (K&S, Section 8.2):

1. **Continuity:** S is a continuous function of its arguments
2. **Monotonicity:** If there are n equal choices ($p_k = 1/n$), then S increases in n
3. **Grouping:** If a choice is broken down into subchoices, S adds according to expectation:

$$S(p_1, p_2, p_3) = S(p_1, p_2 + p_3) + (p_2 + p_3) \cdot S\left(\frac{p_2}{p_2 + p_3}, \frac{p_3}{p_2 + p_3}\right)$$

These are Shannon’s original axioms [7]. K&S shows they follow from the variational framework.

10.4 Formalization

Definition 10.1 (ProbDist). **Lines 19–22, Mettapedia/ProbabilityTheory/Foundations/Distributions/ProbDist.lean**

A probability distribution: probabilities for n outcomes that are non-negative and sum to 1.

Listing 34: Probability distribution

```

1  structure ProbDist (n : N) where
2    p : Fin n → R
3    nonneg : ∀ i, 0 ≤ p i
4    sum_one : sum i, p i = 1

```

Remark 10.2 (KS-facing alias). **Lines 66–72, Information/InformationEntropy.lean**

ProbDist is defined in

Mettapedia/ProbabilityTheory/Foundations/Distributions/ProbDist.lean and re-exported as a KS-facing alias so that Section 8 can refer to ProbDist without importing Foundations directly.

Definition 10.3 (klDivergence). **Lines 279–281, Information/InformationEntropy.lean**

The Kullback-Leibler divergence for probability distributions (K&S Eq. 54).

Listing 35: `klDivergence` (Information/InformationEntropy.lean:279)

```

1 noncomputable def klDivergence {n : N} (P Q : ProbDist n)
2   (hQ_pos : ∀ i, P.p i ≠ 0 → 0 < Q.p i) : R :=
3   sum i, P.p i * log (P.p i / Q.p i)

```

Note: The positivity hypothesis `hQ_pos` ensures Q is strictly positive on the support of P , avoiding $\log(p/0)$ issues. This is the regime where K&S's formula is meaningful.

An extended version `klDivergenceTop` (line 292, Information/InformationEntropy.lean) takes values in $\mathbb{R}_{\geq 0} \cup \{\infty\}$, returning ∞ when this condition fails.

The Shannon entropy is formalized as:

$$S(p) = - \sum_i p_i \log(p_i)$$

with the convention $0 \cdot \log(0) = 0$ (from continuity: $\lim_{x \rightarrow 0^+} x \log x = 0$).

Justified by `zero_mul_log_zero` (line 70).

11 Counterexamples and Clarifications

Directory: Metapedia/ProbabilityTheory/KnuthSkilling/Counterexamples/

11.1 The “Discontinuous Re-grading” Claim

K&S (Section 2) claim that continuity is “merely a convenient convention” and suggest a discontinuous “re-grading” map Θ could preserve the sum rule, using a base-conversion example.

Theorem 11.1 (Re-grading continuity).

File: Counterexamples/RegradeCounterexample.lean

Lean name: regrade_preserving_sum_rule_is_continuous

This claim is false. Any re-grading $\Theta : \mathbb{R} \rightarrow \mathbb{R}$ that preserves:

1. The sum rule: $\Theta(x + y) = \Theta(x) + \Theta(y)$ (additivity)
2. Monotonicity: $x \leq y \Rightarrow \Theta(x) \leq \Theta(y)$

must be linear ($\Theta(x) = c \cdot x$ for some constant c), hence continuous.

Listing 36: Monotone additive functions are linear (RegradeCounterexample.lean:100)

```

1 theorem monotone_additive_is_linear {f : R → R}
2   (hadd : ∀ x y, f (x + y) = f x + f y)
3   (hmono : Monotone f) :
4     ∀ x, f x = f 1 * x
5
6 theorem regrade_preserving_sum_rule_is_continuous {Theta : R → R}
7   (hTheta_add : ∀ x y, Theta (x + y) = Theta x + Theta y)
8   (hTheta_mono : Monotone Theta) :
9     Continuous Theta

```

Remark 11.2 (Why K&S’s Example Fails). K&S’s base-conversion map is not additive: $\Theta(x + y) \neq \Theta(x) + \Theta(y)$. Their example could only work by changing the addition operation to some weird \oplus , which is just obfuscating notation, not demonstrating genuine discontinuity.

The philosophical point (finite systems can’t detect continuity) may be valid, but the mathematical example does not support the claim.

11.2 Pathological Additive Functions

File: Counterexamples/CauchyPathology.lean

Without regularity conditions, Cauchy's equation $f(x + y) = f(x) + f(y)$ has “wild” non-linear solutions (constructed via Hamel bases over \mathbb{Q}) [6]. These solutions are necessarily **non-monotonic**—they oscillate wildly and cannot preserve order.

11.2.1 The Construction

Step 1: Build a Hamel basis (lines 35–77):

We work with \mathbb{R} as a vector space over \mathbb{Q} . First prove $\{1, \sqrt{2}\}$ is \mathbb{Q} -linearly independent (using irrationality of $\sqrt{2}$), then extend to a Hamel basis:

Listing 37: Hamel basis extending $\{1, \sqrt{2}\}$

```

1 theorem linearIndepOn_one_sqrt2 :
2   LinearIndepOn Q id ({(1 : R), R.sqrt 2} : Set R)
3
4 noncomputable def hamelBasis :
5   Module.Basis (...extend {1, sqrt 2}...) Q R :=
6   Module.Basis.extend linearIndepOn_one_sqrt2

```

Step 2: Define the weird map (lines 79–86):

Create a \mathbb{Q} -linear map that sends:

- $1 \mapsto 0$
- $\sqrt{2} \mapsto 1$
- All other basis vectors $\mapsto 0$

Listing 38: Definition of weirdAdditive

```

1 noncomputable def weirdQLinear : R →_[Q] R :=
2   (hamelBasis).constr Q fun i =>
3     if (i : R) = R.sqrt 2 then (1 : R) else 0
4
5 noncomputable def weirdAdditive : R → R := fun x => weirdQLinear x

```

Step 3: Prove it's additive but not linear (lines 87–127):

Listing 39: weirdAdditive satisfies Cauchy's equation but isn't linear

```

1 theorem weirdAdditive_add (x y : R) :
2   weirdAdditive (x + y) = weirdAdditive x + weirdAdditive y
3
4 theorem weirdAdditive_not_mul (A : R) :
5   ∃ x : R, weirdAdditive x ≠ A * x
6   -- Proof: weirdAdditive 1 = 0 but weirdAdditive (sqrt 2) = 1
7   -- So it can't be x ↦ A*x for any constant A

```

Step 4: Convert to positive reals (lines 129–165):

Define $H'(m) := \text{weirdAdditive}(\log m)$ on positive reals:

Listing 40: Multiplicative-additive pathology on positive reals

```

1 noncomputable def Hprime (m : R) : R := weirdAdditive (R.log m)
2

```

```

3 | theorem Hprime_mul (m_x m_y : ℝ) (hx : 0 < m_x) (hy : 0 < m_y) :
4 |   Hprime (m_x * m_y) = Hprime m_x + Hprime m_y
5 |
6 | theorem Hprime_not_B_add_C_log :
7 |   ~ ∃ (B C : ℝ), ∀ m, 0 < m → Hprime m = B + C * log m
8 |   -- Proof: If Hprime m = B + C*log m, then weirdAdditive x = C*x,
9 |   -- contradicting weirdAdditive_not_mul

```

Key insight: These pathological solutions exist but **cannot be monotone**. By the theorem in §11.1, any monotone additive function is linear, so wild solutions like `weirdAdditive` must oscillate wildly and violate order preservation.

11.3 Separation is Not Derivable (Independence Result)

Directory: Metapedia/ProbabilityTheory/KnuthSkilling/Additive/Counterexamples/

A key discovery during formalization: the **separation property is independent** of the base K&S axioms. This was initially unexpected—our early formalization (December 2025) assumed separation was derivable, matching the K&S paper’s informal presentation.

Theorem 11.3 (KS_{Separation} Independence).

File: Counterexamples/KS_{Separation}NotDerivable.lean

Lean name: `exists_knuthskilling_not_separation`

There exists a structure satisfying all base K&S axioms (associativity, strict monotonicity, identity-as-minimum) that fails the separation property.

11.3.1 The Semidirect Product Countermodel

File: Counterexamples/SemidirectNoSeparation.lean

Construction:

$$\begin{aligned} \text{SD} &:= \text{WithBot}(\mathbb{N}^+ \times_{\text{lex}} \mathbb{N}) \\ (u, x) \oplus (v, y) &:= (u + v, x + 2^u \cdot y) \end{aligned}$$

This is a semidirect product: the first component acts on the second via 2^u scaling.

Listing 41: Semidirect product countermodel (SemidirectNoSeparation.lean)

```

1 abbrev SD := WithBot (PN *_[lex] N)
2
3 def baseOp (p q : SDBase) : SDBase :=
4   let u : N := (p.1 : N)
5   (p.1 + q.1, p.2 + (N.pow 2 u) * q.2)
6
7 -- Verified properties:
8 instance : KnuthSkillingAlgebra SD -- All base axioms satisfied
9
10 -- But:
11 theorem SD.not_commutative : ∃ x y : SD, op x y ≠ op y x
12
13 theorem SD.not_KSSeparation : ¬ KSSeparation SD

```

Properties verified:

- ✓ Associative (the semidirect product construction preserves associativity)

- ✓ Strictly monotone in each argument (lexicographic order)
- ✓ Has identity (\perp)
- ✓ Identity is minimum
- ✗ **NOT commutative:** $(1, 1) \oplus (2, 1) \neq (2, 1) \oplus (1, 1)$
- ✗ **Fails KSSeparation:** proven in `not_KSSeparation.lean`

Theorem 11.4 (Independence Theorems).

File: Counterexamples/KSSeparationNotDerivable.lean

```

1 theorem ∃_knuthskilling_noncomm :
2   ∃ (alpha : Type) (_ : KnuthSkillingAlgebra alpha),
3     ∃ x y : alpha, op x y ≠ op y x
4
5 theorem ∃_knuthskilling_not_separation :
6   ∃ (alpha : Type) (_ : KnuthSkillingAlgebra alpha),
7     ¬ KSSeparation alpha

```

Implications:

1. Any proof attempt deriving KSSeparation from the base K&S axioms is **doomed to fail**.
2. Commutativity cannot be derived from the base axioms alone (it requires separation).
3. Separation (or an equivalent like NAP) must be **postulated as an independent axiom**.

Remark 11.5 (Formalization Exposing Hidden Assumptions). This is a concrete example of formalization's value. The K&S paper's presentation suggests separation "follows naturally" from the base axioms. Rigorous formalization revealed this is false: an explicit countermodel blocks all derivation attempts. See Appendix 12.4 for the discovery timeline.

12 Summary: Complete Formalization

12.1 Major Formalized Theorems at a Glance

This section consolidates all the main results proven in the formalization. All paths are relative to `Mettapedia/ProbabilityTheory/KnuthSkilling/`.

Sum Rule (Appendix A)

- `representation_semigroup`
`Additive/Proofs/OrderedSemigroupEmbedding/HolderEmbedding.lean:272`
— $\exists \Theta$ preserving order and addition (identity-free)
- `representation_from_noAnomalousPairs`
`Additive/Proofs/OrderedSemigroupEmbedding/HolderEmbedding.lean:300`
— Full representation with $\Theta(\text{ident}) = 0$
- `associativity_representation`
`Additive/Proofs/GridInduction/Main.lean:54`
— K&S-style (grid/induction) public API
- `op_archimedean_of_separation`
`Additive/Axioms/SandwichSeparation.lean:134`
— Archimedean is derivable from `KSSeparation`

Product Theorem (Appendix B)

- `Psi_is_exp` (`Multiplicative/Main.lean:43`) — $\Psi = \Theta^{-1}$ is exponential under the product equation regularity
- `tensor_coe_eq_mul_div_const` (`Multiplicative/Main.lean:61`) — $x \otimes y = (x \cdot y)/C$
- `ScaledMultRep` (`Multiplicative/ScaledMultRep.lean:44`) — Common interface for both proof paths

Variational (Appendix C)

- `variationalEquation_solution_measurable` (`Variational/Main.lean:310`) — Measurable solution $\Rightarrow H'(m) = B + C \log m$
- `entropyDerivative_variational` (`Variational/Main.lean:290`) — Entropy-derivative satisfies the variational equation
- `entropyForm_deriv` (`Variational/Main.lean:488`) — $\frac{d}{dm} [A + Bm + C(m \log m - m)] = B + C \log m$

Probability Calculus (FOI Mainline)

- `sum_rule`
`Probability/ProbabilityDerivation.lean:856`
— $P(A \vee B) = P(A) + P(B)$ for disjoint events

- `product_rule_ks`
`Probability/ProbabilityDerivation.lean:938`
— $P(A \wedge B) = P(A|B) P(B)$
- `bayes_theorem_ks`
`Probability/ProbabilityDerivation.lean:957`
— Bayes' theorem
- `complement_rule`
`Probability/ProbabilityDerivation.lean:972`
— $P(B) = 1 - P(A)$ for complements

Conditional Probability (Section 7)

- `chainProductRule`
`Probability/ConditionalProbability/Basic.lean:348`
— $p(xy|z) = p(x|z) \cdot p(y|xz)$
- `bayesTheorem`
`Probability/ConditionalProbability/Basic.lean:424`
— $p(x|yz) \cdot p(y|z) = p(y|xz) \cdot p(x|z)$
- `prob_eq_measure_ratio`
`Probability/ConditionalProbability/Basic.lean:714`
— $p(x|y) = m(x \wedge y)/m(y)$
- `baseMeasure_satisfies_measure_axioms`
`Probability/ConditionalProbability/Basic.lean:559`
— m is a probability measure

σ -Additivity Bridge (Extension)

- `ks_sigma_additive` (`Core/ScaleCompleteness.lean:360`) — $\mu(\sup f) = \sum_n \mu(f_n)$ for disjoint countable families

Divergence & Entropy (Sections 6, 8)

- `atomDivergence_nonneg` (`Information/Divergence.lean:102`) — $D(w\|u) \geq 0$
- `atomDivergence_eq_zero_iff` (`Information/Divergence.lean:123`) — $D(w\|u) = 0 \Leftrightarrow w = u$
- `klDivergence` (`Information/InformationEntropy.lean:279`) — KL divergence
- `shannonEntropy` (`Information/InformationEntropy.lean:458`) — $H = -\sum p \log p$

Quantum Theory Classification (Section 4)

- `selection_theorem` (`Mettapedia/Algebra/TwoDimClassification.lean:427`) — $\mu < 0$ gives QM Born rule
- `mean_bornRule_sum_unit_phases` (`Core/SymmetricalFoundation.lean:306`) — Born rule from averaging

- `classification_trichotomy` ([Mettapedia/Algebra/TwoDimClassification.lean:293](#)) — Exactly 3 algebra classes

Counterexamples & Clarifications

- `monotone_additive_is_linear` ([Counterexamples/RegradeCounterexample.lean:100](#)) — Discontinuous re-grading impossible
- `weirdAdditive_add` ([Counterexamples/CauchyPathology.lean:87](#)) — Pathological additive functions exist
- `Hprime_not_B_add_C_log` ([Counterexamples/CauchyPathology.lean:156](#)) — But they violate regularity

12.2 What K&S Claims vs. What We Prove

K&S Claim	Lean Status	Notes
Axioms 0–2 \Rightarrow Sum Rule	Proven	Appendix A representation
Archimedean derivable	Proven	From KSSeparation
Commutativity derivable	Proven	From KSSeparation
Axioms 3–4 \Rightarrow Product Rule	Proven	Appendix B
Variational \Rightarrow Entropy form	Proven	Appendix C
3 algebra classes	Proven	TwoDimClassification
$\mu < 0$ for QM	Proven	selection_theorem
Born rule from averaging	Proven	mean_bornRule_sum_unit_phases

12.3 Key Discoveries from Formalization

1. **Linear order is implicit:** K&S proofs assume trichotomy without stating it.
2. **Identity is essential for positivity:** K&S say the bottom element is “optional” (lines 320, 340–341), claiming “fidelity ensures that other elements are quantified by positive values.” Our formalization **proves this claim is FALSE** for unbounded structures.

The \mathbb{Z} counterexample ([Additive/Counterexamples/NegativeWithoutIdentity.lean](#)):

- $(\mathbb{Z}, +, \leq)$ satisfies K&S Axioms 1–2 (associativity and strict order preservation)
- The Hölder embedding theorem applies (no anomalous pairs)
- But $\Theta(-1) = -1 < 0$ —**negative values appear!**

Why K&S’s claim fails: K&S’s positivity comes from `ident_le`: $\forall x, \perp \leq x$. This requires \perp to *exist* and be *minimal*. For \mathbb{Z} :

- No bottom element exists (\mathbb{Z} is unbounded below)
- The representation theorem still applies
- But positivity is **not guaranteed**

Corrected understanding:

- **With identity + ident_le:** $\Theta(\perp) = 0$ provides canonical normalization; all other elements have $\Theta(x) > 0$.

- **Without identity:** The representation theorem works, but positivity is **not guaranteed**. The embedding is unique up to additive constant, but no constant can rescue positivity for unbounded-below structures.

Status: K&S claim **corrected**. Identity with `ident_le` is **essential** for positivity, not merely “aesthetic.”

3. **Separation property is necessary:** The representation theorem requires an explicit “separation” axiom to enable rational approximation.
4. **Archimedean is derivable:** Not an axiom—follows from separation.
5. **Commutativity is derivable:** Not an axiom—follows from separation via mass counting.
6. **Classification gives isomorphism:** The original K&S classification theorem is about *isomorphism classes*, not equality of multiplication rules.
7. **Measurability replaces continuity:** For Appendix C, measurability (not differentiability) is the correct regularity assumption.
8. **Discontinuous re-grading is impossible:** K&S’s claim that continuity is optional is false for maps preserving both the sum rule and monotonicity.
(Proved in `Counterexamples/RegradeCounterexample.lean`.)
9. **Symmetries 3–4 are product-side:** The product-side symmetries (distributivity, product associativity) are logically separate from the sum-side axioms and are formalized in `Multiplicative/Main.lean`.
10. **Interface design pattern:** Both Appendix A and Appendix B use an **interface + multiple implementations** pattern:

- **Appendix A:** `HasRepresentationTheorem / RepresentationResult` interface with Hölder, cuts, and grid proof paths
- **Appendix B:** `ScaledMultRep` interface with K&S path and Direct path implementations

Downstream code depends only on the interfaces, not on specific proof paths. This separation of concerns allows switching implementations without changing dependent code.

12.4 Sorry Count

Total sorries in core K&S files: 0

All core theorems are fully proven. The formalization is complete and ready for review.

Appendix: Supplementary Derivation Paths (References)

These modules are not needed to audit K&S FOI itself, but they provide additional derivation routes and equivalence theorems that cross-check the entropy/KL story:

- **Unified finite entropy/KL entrypoint:**
`Mettapedia/InformationTheory/EntropyKL.lean`

- **Axiomatic entropy** (Faddeev [12], Shannon [7], Shannon–Khinchin):
[Mettapedia/InformationTheory/ShannonEntropy/Main.lean](#)
- **Shore–Johnson [10] → KL (derivation glue)**:
[Mettapedia/ProbabilityTheory/KnuthSkilling/ShoreJohnson/KLDerivation.lean](#)
 (full import surface: [Mettapedia/ProbabilityTheory/KnuthSkilling/ShoreJohnson/Main.lean](#))
- **Shore–Johnson scope note (Uffink [11])**:
 Uffink critiques the claim that Shore–Johnson “consistency” alone uniquely singles out KL: under weaker formulations (notably weaker forms of system independence), a broader Rényi-family of update rules is compatible. Our Shore–Johnson development therefore treats *product additivity* of the divergence under product distributions as an explicit hypothesis, and derives the KL atom under a standard regularity gate (measurability).
- **Shore–Johnson ↔ Appendix C gate bridge**:
[Mettapedia/ProbabilityTheory/KnuthSkilling/Bridges/ShoreJohnsonVariationalBridge.lean](#)
- **Cox [9] probability calculus derivation**:
[Mettapedia/ProbabilityTheory/Cox.lean](#)
- **Algorithmic probability (Solomonoff [13]) ↔ K&S KL gate**:
[Mettapedia/ProbabilityTheory/KnuthSkilling/Bridges/AlgorithmicProbabilityBridge.lean](#)

Appendix: The KSSeparation Discovery

This appendix documents the discovery that the separation axiom is **independent** of the base K&S axioms—a concrete example of formalization exposing hidden assumptions in informal mathematics.

Timeline

Date	Event	Source
Nov 30, 2025	Initial <code>Algebra.lean</code> created	Commit 53f7a46
Dec 2, 2025	AI agents (likely Claude Code) introduce <code>KSSeparation</code>	Commit af3847d
Dec 4, 2025	1382-line proof attempt (<code>SeparationProof.lean</code>)	Commit d8ed2fb
Dec 19–20, 2025	Ben Goertzel provides “sandwich” framing	v1–v2 documents
Jan 8, 2026	Counterexample proves independence	Commit 6592c37
Jan 13, 2026	<code>SandwichSeparation.lean</code> consolidates results	Commit 9f150fa

Phase 1: Initial Optimism (Dec 2, 2025)

The original comment in `Algebra.lean` was optimistic:

“This is NOT a primitive axiom—it is derivable from the Knuth-Skilling axioms above.
However, for organizational clarity, we factor it into a typeclass...”

This matched the K&S paper’s informal presentation, which treats separation as following naturally from the base axioms.

Phase 2: Failed Derivation Attempts (Dec 4, 2025)

A 1382-line file `SeparationProof.lean` attempted to prove `KSSeparation` from the base axioms. The proof never completed—gaps remained unfilled despite substantial effort.

Phase 3: Ben Goertzel's Contributions (Dec 19–20, 2025)

Ben Goertzel's documents “Foundations of Inference: New Proofs” provided:

- The “sandwich” terminology for the separation property
- Detailed analysis of the A/B/C separation sets
- The “base-indexing” insight (Remark 4, Lemma 7)
- Categorical framing via monoidal functors

Phase 4: The Counterexample (Jan 8, 2026)

The breakthrough: rather than continuing proof attempts, we constructed a **countermodel**. The semidirect product $SD = \text{WithBot}(\mathbb{N}^+ \times_{\text{lex}} \mathbb{N})$ satisfies all base K&S axioms but fails separation.

This formally proved: **no derivation of separation from the base axioms can exist**.

Attribution

- **KSSeparation typeclass:** Introduced by AI agents (likely Claude Code), Dec 2, 2025
- **“Sandwich” terminology & categorical framing:** Ben Goertzel, Dec 19–20, 2025
- **Semidirect product countermodel:** AI agents (likely Claude Code or Codex), Jan 8, 2026
- **Consolidation:** Formalized Jan 13, 2026, combining both contributions

Note: Git commits are authored by the human user (Zar Goertzel), so the specific AI agent cannot be definitively identified from version control. Attribution to “Claude Code” is based on project records showing Claude Code assistance throughout this period.

Mathematical Significance

The countermodel proves that the K&S base axioms (associativity, strict monotonicity, identity-as-minimum) do **NOT** imply:

1. Commutativity
2. The separation property (“sandwich” / rational approximation)

Both must be **postulated as additional axioms** or **derived from stronger assumptions** (e.g., density, Archimedean properties on the real line, or the equivalent No Anomalous Pairs condition from ordered semigroup theory).

Lesson for Formalization

This is exactly what formalization is for: finding gaps that informal proofs skip over. The K&S paper's presentation *suggests* separation follows naturally, but rigorous formalization reveals it requires explicit justification. The countermodel construction took the project from “we can’t prove it” to “we know it’s unprovable.”

13 Build Instructions

From the Mettapedia project root:

```
1 export LAKE_JOBS=3
2 nice -n 19 lake build Mettapedia.ProbabilityTheory.KnuthSkilling
```

For the FOI reviewer entrypoint only:

```
1 export LAKE_JOBS=3
2 nice -n 19 lake build Mettapedia.ProbabilityTheory.KnuthSkilling.
    FoundationsOfInference
```

For memory-intensive grid/induction files (e.g. ThetaPrime.lean):

```
1 ulimit -v 6291456
2 export LAKE_JOBS=1
3 nice -n 19 lake build Mettapedia.ProbabilityTheory.KnuthSkilling.Additive.Proofs.
    GridInduction.Main
```

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