

Exact Execution Verification of Cached GPT-2 in Lean and WebAssembly

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Abstract

This verification requires the compiled neural-network implementation to reproduce the cache and logit bytes specified by the Lean source. We present a checked proof connecting a 19,083-byte WebAssembly artifact to a Lean specification of cached GPT-2 with 124,439,808 parameters and a 128-token context. In the semantics of the WebAssembly interpreter and proof library Talos, written in Lean, the decoded artifact terminates and returns exactly the cache and logit bytes of the Lean token-step recurrence for every weight array of the required length and every valid token sequence of length at most 128. The proof establishes binary decoding, validation, and equality with the execution model, then composes initialization, packed input representation, all twelve transformer blocks, vocabulary projection, allocation sufficiency, and buffer releases. Our Talos extensions define binary32 and binary64 arithmetic through integer operations and explicit rounding. Arithmetic lemmas connect Lean’s logical Float32 operations to Talos for all input words. The standalone WebAssembly runtime Wasmtime executes the module with arithmetic NaN canonicalization enabled. Tests against CPU PyTorch compare 6,432,896 logits across all supported context lengths, with a maximum observed absolute difference of 0.001450, and produce text completions. The final theorem uses three standard logical axioms. Applying the formal result to native execution relies on the interpretation of the formal semantics, the host’s byte I/O, and Wasmtime. The report records the artifact-identity and axiom-policy checks needed to preserve this assurance after future edits.

1 The checked execution result

In the semantics of the WebAssembly interpreter and proof library Talos, written in Lean, the decoded artifact terminates and returns exactly the cache and logit bytes of the Lean token-step recurrence for every weight array of the required length and every valid token sequence of length at most 128. The public theorem is `Project.Gpt2CachedStep.Artifact.artifact_gpt2_128_exact`. It starts from the frozen binary bytes, proves decoding and validation, and establishes the execution specification for their Talos translation [2]. The session proof composes the module’s reset and allocation exports, the mathematical encoding of weight bytes, and successive calls to the cached inference export [8]. It derives the memory conditions needed by every token call from initialization and a uniform allocation budget.

The specification is the executable Lean function `LeanExe.Models.Gpt2.cachedStep`, including its selected FP32 operation order and polynomial numerical routines. LeanExe compiles that function to WebAssembly. The proof uses Talos’s instruction rules for the decoded output [9, 10]. Each arithmetic, tensor, and control-flow result therefore connects to a definition used by the compiler’s input. Per-call postconditions describe returned bytes and heap state, and the session induction preserves the weights and current cache across releases.

This report describes repository revision 4360920c3060d1c625b1860229b5116804d177c8 [1]. The compiler and proof workspace use Lean 4.34.0-rc2 at revision 6a10ac8c22beadecabdbb0919c2b50214762f91d. The Talos revision is 87e3aa5e8f6e6f3b3eb5e7e4c5aba43071002d47. For native execution we use the Bytecode Alliance’s standalone WebAssembly runtime, Wasmtime 44.0.0, with its Cranelift code generator and arithmetic NaN canonicalization enabled [31, 27]. Section 4 specifies the connection between the checked execution model and that runtime.

The implementation loads the pretrained GPT-2 checkpoint as external data. The theorem quantifies over arbitrary bytes of the required weight length, including exceptional FP32 encodings. Consequently, replacing weight values preserves the applicability of the execution proof. The command-line demonstration separately checks the identity of the pretrained checkpoint and rejects a non-finite returned logit before sampling [28]. Those host checks belong to the demonstration’s behavior.

2 Lean algorithm and packed representation

2.1 Model dimensions and interface

The implementation retains the original GPT-2 block organization: pre-normalized attention and feed-forward sublayers, residual additions, final normalization, and a vocabulary projection tied to the token embedding [11]. Table 1 gives the dimensions that determine loop bounds and buffer sizes. The checkpoint contains 1,024 position embeddings, while the implemented cached entry accepts positions zero through 127 [12, 13].

Quantity	Value
Parameter words	124,439,808
Transformer blocks	12
Hidden width	768
Attention heads	12
Channels per head	64
Feed-forward width	3,072
Vocabulary size	50,257
Supported context	128 tokens
Weight payload	497,759,232 bytes
One logit vector	201,028 bytes
Cache increment per token	73,728 bytes

Table 1: Dimensions used by the implementation and proof.

A token is a vocabulary index represented by `UInt32`. Context length counts these indices. Weights, activations, cached keys and values, and logits use little-endian packed 32-bit words in Lean `ByteArray` values. The source operations `word` and `generateUInt32LE` read and construct those words. The compiler emits scalar FP32 instructions and packed memory operations for them [14, 15].

The source entry has the type

`ByteArray → ByteArray → UInt32 → Nat → CachedResult.`

Its arguments are weights, the prior cache, a token, and its position. A `CachedResult` contains the extended cache and the next-token logit vector. At position p , the entry checks weight length,

token range, $p < 128$, and cache length $73,728p$. A failed check returns two empty byte arrays. The separate rejected-entry theorem proves four zero result words and complete store preservation under the public entry’s representation conditions [16].

The host API passes the weight pointer and byte length, cache pointer and byte length, token, and position, in that order. The export returns cache pointer, cache byte length, logit pointer, and logit byte length. Talos represents call arguments and results in operand-stack order, which reverses these lists. Checked equalities identify `cachedStep`, `alloc`, `reset`, and `release` with function indices 38, 39, 40, and 42 [17].

2.2 Cached attention and token recurrence

The cache stores both 768-component key and value vectors for each layer and prior token. Within a position, the twelve layers occur in increasing order, with keys preceding values in each layer. If s is a prior position, ℓ is a layer, and $o < 1536$ selects a key or value component, its word index is

$$((12s + \ell)1536) + o.$$

For the current position, `cachedKv` reads the key or value from the newly computed QKV projection. The source calculates attention scores by a 64-term ordered dot product, scales by $1/8$, normalizes over positions zero through the current position, and computes the weighted value sum [13].

Each block returns a new hidden vector and its current key/value vectors. The enclosing twelve-layer loop concatenates those key/value vectors and appends them to the old cache. Token and position embedding precede that loop. Final layer normalization and the tied 50,257-row vocabulary projection follow it. The cached Lean function specifies this entire token step. The formal target takes its recurrence as the algorithm. Comparison with the separate full-prefix implementation appears as a runtime test in Section 5.

Let W be a weight byte array and $t_0, \dots, t_{n-1}$ the input tokens. Define

$$C_0 = \epsilon, \quad (C_{p+1}, L_p) = \text{cachedStep}(W, C_p, t_p, p), \quad 0 \leq p < n, \quad (1)$$

where ϵ is the empty byte array and L_p is the logit byte array. The Lean definition `Session.sourceTrace` returns the list of these cache/logit pairs. Every equality to this trace is byte equality, so it includes signed zeros and the numerical routines’ exceptional outputs.

2.3 Arithmetic order and numerical routines

Matrix projections accumulate each output component in source loop order, with a separate FP32 multiplication and addition at each step, followed by bias addition. Layer normalization computes an ordered mean, an ordered sum of squared deviations, division by 768, addition of the FP32 word `0x3727C5AC` for epsilon, square root, and reciprocal. It then applies the centered value, inverse standard deviation, scale, and bias in the written order [14]. These choices determine the exact source result.

The source implements exponential evaluation by a degree-18 Horner polynomial with explicit coefficient words. For finite nonpositive arguments, `expNeg` returns zero below -64 , halves an argument below -1 at most six times, evaluates the polynomial, and squares the result once per halving. GELU uses an exponential form of the tanh approximation and a magnitude cutoff at eight. Beyond that cutoff, the finite positive tail returns the input and the finite negative tail returns zero [18]. The bitwise comparisons and branch definitions in the cited source determine behavior on all other input words. The execution proof includes those branches.

The polynomial, constants, cutoff tests, and rounding order form part of the algorithm specification. The PyTorch reference uses its own tensor and transcendental implementations. The report’s numerical comparisons measure their agreement on the recorded checkpoint and token sequences. Proving a real-valued approximation bound would require a separate statement relating these numerical routines to real formulas.

3 Statement and construction of the proof

3.1 Termination and observable results

For a fixed Talos module M , environment E , function index i , initial store S , argument stack a , and postcondition P , write $\text{TW}(E, M, i, S, a, P)$ for Talos’s **TerminatesWith**. Its definition is

$$\exists N \forall f \geq N \exists v, S'. \quad \text{run}(f, M, i, S, a, E) = \text{Success}(v, S') \wedge P(S', v).$$

The natural number f is interpreter fuel. This predicate proves eventual successful execution and the postcondition for every sufficiently large fuel value [19]. The generated GPT-2 module has no imported functions, and the theorem quantifies over the host environment.

Theorem 1 (Cached GPT-2 artifact execution). *Let B be the frozen 19,083-byte artifact at the report’s source revision. Decoding and validation succeed for B . Its decoded module satisfies the formal binary grammar and validation rules of the supported profile. Let M be its Talos translation, W a byte array of length 497,759,232, and $T = [t_0, \dots, t_{n-1}]$ a list of **UInt32** tokens with $n \leq 128$ and $t_p < 50,257$ for every $p < n$. Starting from M ’s initial store, reset and weight allocation terminate successfully. After the specified byte copy of W into the allocated payload, the token calls and releases described below terminate successfully. At each position p , their observable cache and logit bytes equal C_{p+1} and L_p in Equation (1).*

The checked declaration for Theorem 1 is `Project.Gpt2CachedStep.Artifact.artifact_gpt2_128_exact` [2]. Its execution component transfers `Project.Gpt2CachedStep.Spec.gpt2_128_exact` to the decoded module [8]. The three input conditions in the statement suffice, including for the empty token list. For a nonempty list, `Session.RunsFor` calls `cachedStep`, observes the new cache, releases the previous cache when its pointer is nonzero, observes the logits, releases the logit buffer, and continues with the returned cache. The step theorem preserves ownership of the weights and current cache through those releases. The source trace specifies the outputs of this call sequence.

The weight-copy step is the pure store update `PackedInput.write`. Its theorem establishes the required byte representation and heap ownership after allocation [20]. This specifies the byte I/O boundary used by the execution theorem. The native host performs the corresponding copy with `memcpy`. The proof’s trust boundary includes that implementation.

3.2 Floating-point extensions to Talos

Our extensions give Talos executable `binary32` and `binary64` arithmetic definitions over integers and raw words. The earlier evaluator stored floating-point values as `UInt32` and `UInt64` words, then called Lean’s native `Float32` and `Float` operations to evaluate arithmetic [32]. We added the `Wasm.IEEE32` and `Wasm.IEEE64` definitions and changed the interpreter’s arithmetic dispatch to use them [33, 34]. These definitions expose the arithmetic computation to Lean proofs.

The models decode sign, exponent, and fraction fields. Finite `binary32` values have the form $z 2^{-149}$ for an integer z , and finite `binary64` values have the form $z 2^{-1074}$. Addition forms the exact

integer sum at that scale. Multiplication forms an exact dyadic product, and division forms an exact rational quotient. Output rounding selects the nearest representable value and resolves ties toward an even significand. The square-root implementation compares the exact radicand with a squared halfway point to choose the rounded result [33]. The integer definitions include subnormal values, gradual underflow, overflow, signed zeros, infinities, and canonical arithmetic NaNs.

The extension also gives integer definitions for comparisons, minimum and maximum, sign operations, rounding to an integer, and signed and unsigned integer conversions. Trapping and saturating conversions have explicit range and exceptional-value branches. Sign operations preserve the other word bits, including a NaN payload. The big-step evaluator, small-step evaluator, and weakest-precondition rules use the corresponding arithmetic definitions and NaN classifiers [34, 35]. At the pinned revision, cross-format promotion and demotion use Lean’s conversion functions. The generated GPT-2 module uses scalar binary32 arithmetic and raw-word reinterpretation.

We added operation theorems to Talos’s CodeLib proof library for finite arithmetic, exceptional values, comparisons, rounding, and conversions. For example, in namespace `CodeLib.IEEE32`, `mul_finite_rounder` identifies the product word with the dyadic rounder, and `sqrt_signed_zero` establishes the sign-preserving zero case. Total execution theorems connect selected instruction sequences to these word functions, including multiplication, division, square root, conversion, and SIMD examples [35]. The extension replaced the legacy `IEEE32Exec` compatibility axioms with proved equalities over the integer definitions [36].

The extended library also contains real-error bounds under stated input conditions and proofs that compose rounding errors through Horner evaluation and dot products [37]. Those results support numerical analyses of floating-point algorithms. The GPT-2 proof uses the exact operation definitions and execution rules, together with the source-correspondence equalities in the next subsection. Talos supplies the arithmetic semantics. The LeanExe proof workspace proves equality with Lean’s logical `Float32` definitions for the five GPT-2 arithmetic operations on every input word.

3.3 Arithmetic correspondence

For each binary operation $\circ \in \{+, -, \times, /\}$, the arithmetic proof establishes

$$\forall a, b : \text{UInt32}, \quad \text{LeanExe.Float32.opBits}(a, b) = \text{Wasm.IEEE32.op}(a, b).$$

An analogous theorem covers square root. Here `opBits` and `op` stand for the corresponding named operation. The source intrinsics reduce to Lean’s logical `Float32.Model`, and the proofs compare its encoding and rounding definitions with Talos’s `IEEE32` operations [21, 22].

The proofs split exceptional values from finite values, including zero and subnormal cases. Shared lemmas relate encodings, significands, exponents, scaling, and rounding. Finite multiplication, division, and square root require their respective rounding arguments. NaN and infinity cases close against the explicit logical definitions. The resulting equalities quantify over every input word and supply the scalar facts used by the tensor-loop proofs. The formal source meaning comes from Lean’s logical definitions. Native Lean evaluation appears in separate execution tests.

3.4 Tensor loops and ownership

The packed-memory proofs connect emitted loads and stores to source byte-array reads and construction. Loop invariants relate the constructed memory prefix or scalar accumulator to the source recurrence, while frame conditions preserve protected input regions. A shared range-loop

rule supports an arbitrary preserved operand-stack prefix, which the generated matrix projection needs because it retains a destination address during dot-product evaluation [23].

Kernel theorems compose into normalization, cached attention, complete transformer blocks, the twelve-layer hidden traversal, and vocabulary projection. Table 2 identifies the main composition boundaries. The accepted-entry theorem returns both output buffers’ ownership, their separation from each other, and freshness relative to the initial heap [16]. These properties permit the session theorem to free the previous cache and the current logits while preserving the weights and new cache.

Proof boundary	Established result
Packed access and construction	Source word values, initialized output prefix, address conditions, and preservation.
Scalar FP32 kernels	Ordered reductions, polynomial evaluation, comparisons, and activation results.
Cached attention	Cache lookup, score normalization, weighted values, allocations, and temporary releases.
Transformer block	Both hidden and key/value outputs, ten tensor-kernel calls, and cleanup.
Hidden traversal	Embedding, twelve blocks, cache assembly, ownership, and cleanup.
Vocabulary and entry	All 50,257 output words, accepted and rejected branches, and protected inputs.
Initialization and session	Encoded input, sufficient allocation capacity, and successive calls and releases.

Table 2: Composition of the checked execution proof. The main proof import identifies the component declarations [16].

3.5 Allocation sufficiency through position 127

The allocation proof bounds heap-top growth by charging each requested allocation, including alignment and allocator overhead. A successful free-list reuse preserves the heap top, as does release. Thus the argument applies to the existing allocator independently of which reusable block it selects [24].

Let H_p be the heap top before token call p , and set

$$B_0 = 536,870,912, \quad D = 16,777,216.$$

These are 512 MiB and 16 MiB, respectively. Initialization proves $H_0 \leq B_0$. The session invariant contains

$$H_p \leq B_0 + pD. \tag{2}$$

For $p < 128$ and a cache of length $73,728p$, **Entry.budget** proves every allocation-capacity premise and bounds the resulting heap top by $H_p + D$, provided $H_p + D < 2^{32}$. Equation (2) supplies that premise because

$$B_0 + 128D = 2,684,354,560 < 2^{32}.$$

The session induction therefore closes all 128 positions [24, 8].

The memory invariant requires a 65,536-page capacity and bounds the page count by that capacity. Each page contains 65,536 bytes, giving a 4 GiB memory bound. The 2.5 GiB expression above bounds the charged heap top. The measured linear-memory size in Section 5 is a separate observation. Host allocation success within the modeled capacity and sufficient native execution resources belong to the runtime assumptions.

4 Artifact and runtime boundary

4.1 Binary decoding and execution transfer

The artifact proof embeds all 19,083 bytes as the Lean value `artifactBytes`. Checked parser certificates establish successful decoding of each function body, section, and the complete file. The decoder soundness theorem connects that result to the declarative binary grammar. Successful validation then yields `CoreValid`, which states section, memory-limit, global, export, and function-typing conditions for the supported binary profile [3, 2].

The translation proof identifies each of the 43 decoded function bodies with the corresponding function in the execution model. A whole-module equality also identifies types, exports, globals, memory, and the remaining module fields. Equality substitution transfers the module-parameterized session specification `Spec.ExactSpecFor` to the decoded module [2, 8]. With $B = \text{artifactBytes}$, the final theorem has the form

$$\begin{aligned} \exists R, V. \quad & \text{decode}(B) = \text{ok}(R) \wedge \text{Grammar}.\text{Encodes}(B, R) \\ & \wedge \text{validate}(R) = \text{ok}(V) \wedge \text{CoreValid}(R) \\ & \wedge \text{ExactSpecFor}(V.\text{toTalos}). \end{aligned}$$

Here R is the raw decoded module, V is its validated representation, `Grammar.Encodes` is the declarative encoding relation, and `toTalos` is the formal translation. `ExactSpecFor` states the initialization and session behavior in Theorem 1. The compiler’s output is the fixed subject of this proof.

The frozen artifact and the reviewed runtime binary have SHA-256

e93de126e00d7f5c5b9b30ca014a13b1385e9f91e3cb6b4e56a4aacf7a2b4ade.

The package checker compares the external file with both the frozen package and the embedded byte array. It then builds the artifact and behavior proofs, checks the registered artifact statements, and audits the registered declarations’ axiom dependencies [4]. A byte-for-byte comparison also confirmed the identity of the current command-line binary during the statement review [5]. File reading and comparison connect the mathematical byte array to the file used for execution.

The source-driven check remains available: it recompiles `cachedStep`, renders the binary as WAT, generates a Talos module, and compares that module with the tracked execution model [25]. The recorded focused source check and exact-artifact check passed using the existing build cache [5]. The runtime measurements in Section 5 retain the saved 18 September canonical-mode test record.

4.2 Logical foundations and statement review

The Lean kernel checks the proof terms. The final artifact theorem’s axiom report contains exactly `propext`, `Classical.choice`, and `Quot.sound`. Propositional extensionality identifies logically equivalent propositions. Choice selects an element from a type whose nonemptiness has been proved and supports classical reasoning. Quotient soundness identifies quotient values represented by related elements. These three principles supply the mathematical foundations used by the proof dependencies [7, 5].

Native-evaluation axioms accept the result of a compiled computation and therefore add trust in the compiler and execution machinery used for that computation. The generic package policy accepts these additional assumptions. The statement review audited the final GPT-2 theorem against the three mathematical axioms [4, 5].

Seven checked review lemmas test consequences of the public statement [6]. They exhibit weights and a 128-token list satisfying its premises, transfer the specification to every successful decoding and validation of the frozen bytes, and identify the decoded exports used by the proof. They establish one source result per token, the required cache size after a valid step, and 50,257 FP32 logits in every result. The first-token lemma derives explicit successful `Wasm.run` witnesses for reset, allocation, and inference from the final artifact theorem, with the required result lengths and exact source cache and logit bytes in memory. All seven lemmas pass with the same three-axiom policy or a subset. The full token-list induction derives the memory and cache conditions through position 127 [8, 6].

4.3 Native execution and automated checks

WebAssembly permits multiple NaN results for some floating-point operations [26]. Lean and Talos choose a canonical arithmetic NaN word, `0x7FC00000` for binary32. The C host selects `Cranelift` and enables `wasmtime_config_cranelift_nan_canonicalization_set` [27, 28]. Wasmtime’s implementation of that mode remains trusted. The recorded FP32 tests check its exact NaN words, including signaling inputs and negative or noncanonical payloads [29]. The selected runtime mode is part of the deployment boundary for exact raw-word agreement.

The formal binary grammar, validation rules, translation, and Talos execution definitions specify the modeled WebAssembly behavior. Applying their result to a native run trusts the correspondence of those definitions with WebAssembly, Wasmtime and its native-code backend, the C host, and the host’s byte I/O. Tokenization and token selection occur outside the proved recurrence. The theorem accepts vocabulary indices and produces logit bytes, which lets either a user-selected sequence or a generated sequence supply the next valid token.

Two automated checks remain open at the report checkpoint [5]. The CLI recompiles the current source and executes the result. Enforcing equality with the registered artifact is a manual responsibility after a compiler or source edit. The current compiled file passed that comparison. The package gate audits the manifest’s artifact-equality and behavior declarations, while the combined `artifact_gpt2_128_exact` theorem receives a separate review audit. The package checker’s axiom policy permits native-evaluation assumptions used by other packages. Continuing the reviewed assurance requires a CLI artifact-identity check and an automated audit of the combined GPT-2 statement restricted to its three mathematical axioms [28, 4].

The theorem fixes the module’s architecture and the 128-token limit and quantifies over weight arrays and token lists satisfying the stated conditions. A larger architecture or context needs corresponding source, generated-module, and proof updates. The scalar arithmetic and packed-memory lemmas accept inputs beyond this one model. The composed transformer and allocation theorems retain their stated dimension bounds. Numerical accuracy against real arithmetic, equivalence between the cached and separate full-prefix Lean algorithms, and language-quality properties remain separate obligations.

5 PyTorch comparisons and command-line inference

The runtime tests use the pinned pretrained checkpoint with CPU PyTorch 2.9.1 and Transformers 4.57.6. The manifest checks model files by hash and records checkpoint revision `607a30d783dfa663caf39e06633721c8d4cfc7e` [12]. The CPU reference uses FP32, eager attention, and one PyTorch thread. The WASM driver keeps weights and cached key/value data in a resident Wasmtime instance [28].

The 128-position test starts with the story prompt “Once upon a time, in a small village” and appends repetitions of a fixed sentence before taking the first 128 token IDs. For each prefix length from one to 128, it compares all 50,257 cached WASM logits with the corresponding CPU PyTorch logits [30]. This gives $128 \times 50,257 = 6,432,896$ comparisons. Every component satisfies the test tolerance

$$|\ell_{\text{wasm}} - \ell_{\text{torch}}| \leq 0.002 + 10^{-4} |\ell_{\text{torch}}|.$$

The maximum observed absolute difference is approximately 0.001450 [29]. The quantified exact-equality theorem concerns the Lean source, while this tolerance test concerns the checkpoint and sequence just described.

The canonical-mode run took 100.2 seconds and reached approximately 1.03 GiB of WASM linear memory. Its allocation counters recorded 33,025 allocations and 33,023 frees, leaving the weights and current cache. The allocator reuses whole freed blocks. The final cache contains $128 \times 73,728 = 9,437,184$ payload bytes. The test also checks the four rejected-input conditions, cache reset and replay, and byte-exact agreement between cached and full-prefix WASM logits for the nine-token story prompt [29, 30]. These timings describe the recorded local runs and support execution feasibility. They provide no cross-system speed comparison.

Three recorded completions generated 64, 32, and 16 tokens in 55.6, 27.8, and 15.4 seconds. The 16-token greedy completion matched PyTorch’s token sequence. The 64-token sampled completion began:

Once upon a time, in a small village called Hukur. The village is a great expanse of white sand and the winds blow fast

The saved test record includes the complete strings and generation settings [29]. They demonstrate the command-line path from tokenizer output through the resident WASM module and back to text.

From a prepared repository checkout, the WASM and PyTorch commands are:

```
tools/gpt2 --text 'Once upon a time' --generate 32 --seed 42
tools/gpt2-pytorch --text 'Once upon a time' --generate 32 --seed 42
```

Both use a shared uv project with pinned Python dependencies. The default generation settings are top- k 40, temperature 0.8, and seed 42. The WASM client uses the Lean SplitMix64 generator for random draws and Python for sampling probabilities. PyTorch uses its own seeded generator, so equal seeds produce different draw sequences. Adding `--top-k 1` to each command selects greedy decoding. Small logit differences can still change an argmax [28].

The setup command

```
uv run --project training/gpt2 training/gpt2/reference.py fetch
```

downloads and checks the checkpoint. The reproduction guide lists the compiler and Wasmtime prerequisites [38]. After a WASM command has produced `build/gpt2-124m/inference/cachedStep.wasm`, the focused artifact check is

```
tools/artifact-proof.js check \
  build/gpt2-124m/inference/cachedStep.wasm \
  Project.Gpt2CachedStep.ArtifactTranslation
```

This check enforces binary identity as well as the registered proof checks. The separate statement review and final-theorem axiom audit run with

```
tools/leanrun --timeout 3m lake -d proofs/talos/lean env lean \
  proofs/talos/reviews/gpt2-2026-09-19.lean
```

The source-regeneration and numerical-comparison commands are

```
tools/talos-proof.js check gpt2_cached_step
node test/packed.js --gpt2-cached --gpt2-completions
```

6 Proof reuse and verification scope

The construction isolates scalar arithmetic, packed representation, loop correspondence, and allocation accounting into reusable lemmas. The final session proof uses an induction on the token list, with one step theorem carrying output ownership into the next call. Its numerical inputs range over arbitrary weight bits, so the proof’s size depends on algorithm structure and dimensions rather than checkpoint values. The heap budget uses allocation sizes to close the concrete invocation premises without inspecting numerical weights.

The earlier LeanExe subset report describes the compiler’s accepted types, packed layouts, and per-compilation correspondence obligation [9]. The present result connects the frozen binary bytes to the cached GPT-2 Lean algorithm under Talos semantics and the stated byte I/O boundary. The original GPT-2 source supplies the architecture and cached-attention organization [11]. The checked Lean algorithm fixes the FP32 numerical choices used here, and the PyTorch tests provide checkpoint-specific comparison evidence. Further verification of native execution would address Wasmtime, its code generator, and the host’s implementation of the byte I/O boundary.

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