

PastaLean

Transpiling and *Verifying* Python in Lean 4 —
without LLMs

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Organized by **IISc Bengaluru** × **Emergence AI**

[GitHub](#) [PyAstLean](#)

Overview

- **Why** verify Python at all — and why in Lean

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- **How** PastaLean turns Python into Lean, node by node
- **What** we support: control flow, classes, libraries...
- **Proving** the generated code correct

What are we cooking?

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Pasta

What are we cooking?



Pasta

+

What are we cooking?



Pasta

+



Python

What are we cooking?



Pasta

+



Python

+

PastaLean

What are we cooking?



Pasta

+



Python

+



Lean 4

PastaLean

What are we cooking?



Pasta

+



Python

+



Lean 4

=

PastaLean

What are we cooking?



Pasta

+



python™

Python

+

LEAN

Lean 4

=

PASTA

PastaLean

What are we cooking?



Pasta

+



Python

+



Lean 4

=



~~P~~·~~ast~~·~~a~~Lean → PastaLean

PastaLean

What are we cooking?



Pasta

+



Python

+



Lean 4

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P·ast·aLean → PastaLean

How to eat(use) PastaLean?

PastaLean

What are we cooking?



P·ast·aLean → PastaLean

How to eat(use) PastaLean?

- Take ordinary, *logical* Python — the kind data scientists and practical logical code's LLMs actually write.

PastaLean

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- Preprocess Python code for Type annotations, obtain it's Abstract Syntax Tree(AST) and give Lean it's JSON format.

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The Goal: write in Python, get the guarantees of a proof assistant.

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- We use **Lean 4**: a proof assistant that is *also* a fast, real programming language.

PastaLean

Why Lean?

Lean 4 is a proof assistant *and* a general-purpose language (it compiles to C — genuinely fast).

1.9M+

lines in Mathlib, its math library

282k+

theorems · 134k definitions

772

contributors · 1 tiny trusted kernel

- **Powerful automation** — tactics like `simp`, `grind`, `aesop`, etc. search proofs and close goals for you, often in one line.
- **Endlessly customizable** — custom syntax, macros, and tactics let you extend the language itself — exactly how PastaLean is built.
- **Trusted where it matters** — used by Fields medallist **Terence Tao**, backed by the **Lean FRO**, applied to industrial verification.
- A small, auditable **kernel** re-checks every proof — nothing is taken on faith.

See Leo de Moura's blog [Why Lean?](#).

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And because pure Python becomes bare Lean *terms*, we can state and prove theorems about it.

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Prompt an LLM to translate	PastaLean (AST transpiler)
Hallucinates; silently changes semantics	Deterministic — same node, same Lean, always
Not reproducible, not auditable	One runtime pins <i>what each operation means</i>
Breaks on long / scaled programs	Compositional: per-statement, streamed
No ground truth that it is correct	Output is <i>checked</i> by the Lean elaborator

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Other Possible methods maybe?

- Other Methods like using FFI is possible, however theorem proving would not be possible.
- Converting Lean to Python is possible, however legacy python checking would not be possible and there are a lot more python programmers.

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```
def sum_to_n := fun (n : Int) ↪
  (do
    let mut total := (0 : Int)
    for i in (PastaLean.pyRange (n + 1 : Int))do
      let _ := Libraries.passta.pyPassInvariant ((2 : Int) * i == total)
      total := total + i
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def sum_to_n := fun (n : Int) ↪
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        total := total +p i
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↓ *prove — automatically*

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theorem sum_to_n_spec : ⌈p n ≥ (0 : Int)⌋p sum_to_n n ⌋p ⇒ ⌈p
  by
  mvcgen [sum_to_n, PastaLean.pyRange_forIn, PastaLean.pyRange_forIn]
  · ⌋p {cur, total} ⇒
    ⌈p let i := (cur.prefix.length : Int);
      (2 : Int) *p total = i *p (i -p (1 : Int))⌋p
  simp_all [taste_ingr]; grind +locals +suggestions
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```
#eval sum_to_n 10
```

55

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Difference	Provable	Computable
Python <code>float</code>	Converted to $\mathbb{Q}$ or $\mathbb{R}$ (if transcendental functions are used).	Lean's inbuilt <code>Float</code> is always used.
Provable	Yes	Sometimes depending on function, not always.
Computable	Sometimes. If $\mathbb{R}$ is used then never.	Yes
API differences(<code>print</code> , <code>exceptional-handling</code>)	Special non-IO Monad based functions are used to support proving	IO Monad functions are used for real behaviour

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def euclidean_distance(p1, p2):  
    if len(p1) ≠ len(p2):  
        raise ValueError("Points must have the same number of dimensions")  
  
    # Using zip, list comprehension, and math.pow  
    sq_diffs = [math.pow(a - b, 2) for a, b in zip(p1, p2)]  
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noncomputable def euclidean_distance := fun (p1 : List Int) ↦ fun  
  ((do  
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      throw  
      (PastaLean.PyException.Raise "ValueError"  
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    else  
      let _ := ()  
      -- Using zip, list comprehension, and math.pow  
      let mut sq_diffs :=  
        (PastaLean.pyIter (PastaLean.pyZip p1 p2)).map fun _pair_  
          let a := Prod.fst _pair_1;  
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          Libraries.math.pyMathPowExact (a -p b) (2 : Int)  
      let __py_ret_1 := Libraries.math.pyMathSqrtR (PastaLean.pyS  
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        (PastaLean.pyIter (PastaLean.pyZip p1 p2)).map fun _pair_  
          let a := Prod.fst _pair_1;  
          let b := Prod.snd _pair_1;  
          Libraries.math.pyMathPowExact (a -p b) (2 : Int)  
      let __py_ret_1 := Libraries.math.pyMathSqrtR (PastaLean.pyS  
      return __py_ret_1) :  
    PastaLean.PyExceptId _)
```

```
def euclidean_distance'rn : List Int → List Int → PastaLean.PyExc  
  fun (p2 : List Int) → do  
    if h_1 : PastaLean.pyLen p1 ≠ PastaLean.pyLen p2 then  
      throw  
      (PastaLean.PyException.Raise "ValueError" (ToString.toStr  
    else  
      let _ := ()  
      -- Using zip, list comprehension, and math.pow  
      let mut sq_diffs :=  
        (PastaLean.pyIter (PastaLean.pyZip p1 p2)).map fun _pair_1 ⇒  
          let a := Prod.fst _pair_1;  
          let b := Prod.snd _pair_1;  
          Libraries.math.pyMathPow (a -p b) (2 : Int)  
      let __py_ret_1 := Libraries.math.pyMathSqrt (PastaLean.pySum sq  
      return __py_ret_1
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Functions	<code>def</code> , <code>lambda</code> , <code>return</code> , default args, nested defs
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Branches & recursion

Loops & mutation → an `Id.run do` block:

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Branches & recursion

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def fibonacci(n: int):  
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def classify(nums: list[int]):  
    total = 0  
    for n in nums:  
        if n % 2 = 0:  
            total += n  
        else:  
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Statements outside functions(in open) are also well supported.

Data structures & comprehensions

`list` · `dict` · `set` · `tuple`, slicing, indexing, container methods:

Comprehensions → `map` / `filter`

List & string ops

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squares = [x*x for x in range(10)
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cubes = {n: n**3 for n in range(5)}

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  (List.filter (fun x => x % 2 == 0)
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`x[i]` → `pyGetItem`, slicing → `pySlice`, `iter` → `pyIter`

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        let _ ← pyPrintNoop [pyPrintArg "Division by zero error"]  
        let __py_ret_1 := -(1 : Int)  
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      else  
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    finally  
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PyExceptId (pure, provable) vs PyExcept = ExceptT PyException IO (real IO).

Classes / OOP

A Python `class` becomes a Lean `structure` plus namespaced method definitions. `self` is the structure value, and mutation is value-semantics (`C.new` builds one):

```
class Point:
    def __init__(self, x,
                  self.x = x
                  self.y = y

    def norm_sq(self):
        return self.x*self.x
               + self.y*self.y
```

```
obj = Point(1,2)
norm = obj.norm_sq()
```

```
structure Point where
  x : Int
  y : Int
  deriving Inhabited, Repr, BEq

def Point.new := fun x ↦ fun y ↦ ({ x := x, y := y } : Point)
def Point.norm_sq := fun (self : Point) ↦ self.x *p self.x +p self.y *p self.y

def obj := Point.new (1 : Int) (2 : Int)
def norm := Point.norm_sq obj
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```
#check obj
| obj : Point

#eval norm
| 5
```


PastaLean

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- All you do is write the Lean function for your Python implementation, use our builtin typeclasses support if needed. All Libraries are independently implementable.
- Now just map the member name to a Lean runtime function. This will automatically integrate with our code generation pipeline.
- `numpy` , `pandas` , `scipy` , `math` , `typing` ship today; anything unsupported degrades to a flagged `pyUnsupported` so the file still compiles.

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where `pyNumpyNorm` and other functions are as simple as:

```
def pyNumpyNorm {α} [PyNumpyScalar α] (xs : List α) : Float :=
  let ys := xs.map toFloat
  Float.sqrt (ys.foldl (fun acc x => acc + x * x) 0.0)
```

PastaLean

Showcases

Real programs — transpiled, type-checked, and (for the pure parts) *proved*. Full code on [GitHub](#). Press ↓.

Showcase	Python	→ Lean	Checks in	Exercises
<code>eg1</code>	39 loc	148 loc	5.2 s	exceptions · <code>zip</code> · comprehensions · <code>math</code>
<code>eg2</code>	36 loc	97 loc	5.1 s	<code>numpy</code> mean · matmul · <code>try / except</code>
<code>orbital</code>	186 loc	386 loc	7.8 s	vector algebra · conserved quantities · <i>proved</i>
<code>cnn</code>	141 loc	355 loc	5.4 s	a from-scratch CNN class · forward + backprop
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Orbital Physics · conserved quantities

```
def cross_x(ax: float, ay: float, az: float, bx: float, by:
    """x-component of a x b."""
    return ay * bz - az * by

def norm_sq(ax, ay, az):
    return ax*ax + ay*ay + az*az

def kinetic(m, vx, vy, vz):
    return 0.5 * m * norm_sq(vx, vy, vz)

def momentum(m, vx, vy, vz):
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def spring_energy(k: float, dx: float, dy: float, dz: float
    """Hooke potential energy (1/2) k |d|^2 stored in a spr
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def momentum_conserved(m1: float, v1: float, m2: float, v2:
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    assert (m1 * v1 + j) + (m2 * v2 - j) == m1 * v1 + m2 *

def spring_force_is_central(k: float, dx: float, dy: float,
    """The Hooke force  $F = -k d$  is central (anti-parallel t
    no torque about the spring axis:  $d \times F = 0$  (x-component
    assert cross(d, dx, dy, dz) == 0
    return True
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  ay *p bz -p az *p by

def norm_sq := fun (ax : Rat) → fun (ay : Rat) → fun (az : Rat) → ax *p ax +p ay *p ay +p az *p az

def kinetic := fun (m : Rat) → fun (vx : Rat) → fun (vy : Rat) → fun (vz : Rat) → (0.5 : Rat) *p m *p norm_sq vx vy vz

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  -/
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@[taste_ingr]
theorem momentum_conserved :
  ∀ (m1 : Rat),
    ∀ (v1 : Rat), ∀ (m2 : Rat), ∀ (v2 : Rat), ∀ (j : Rat), m1 *p v1 +p j +p (m2 *p v2 -p j) = m1 *p v1 +p m2 *p v2
  by intros; simp_all (config := { zetaDelta := true }) [taste_ingr]

@[taste_ingr]
theorem spring_force_is_central :
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  by intros; simp_all (config := { zetaDelta := true }) [taste_ingr]; grind +locals
```


cnn · a CNN, transpiled

```
class CNN:
    def conv(self, img):
        # valid 2x2 conv: 8x8 → 7x7
        out = []
        for i in range(7):
            row = []
            for j in range(7):
                s = 0.0
                for a in range(2):
                    for b in range(2):
                        s += img[i+a][j+b] * self.kernels[a][b]
                row.append(s)
            out.append(row)
        return out
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      out.append(row)
    return out
```



```
structure CNN where
  kernel : List (List Real)
  dense_w : List Real
  dense_b : Real
  deriving Inhabited

noncomputable def CNN.conv := fun (self : CNN) => fun (img : List (List Rat)) =>
  Id.run
  (do
    let mut out := []
    for i in (PastaLean.pyRange (7 : Int))do
      let mut row := []
      for j in (PastaLean.pyRange (7 : Int))do
        let mut s := (0.0 : Real)
        for a in (PastaLean.pyRange (2 : Int))do
          for b in (PastaLean.pyRange (2 : Int))do
            s := s +_p img[i +_p a][j +_p b] *_p self.kernel[a][b]
          row := PastaLean.pyAppend row s
        out := PastaLean.pyAppend out row
      return out)
```



```
def process_data(data, weights):  
    try:  
        m = np.mean(data)  
        print(f"Mean: {m}")  
        centered = np.subtract(data,  
                                [[m, m], [m, m]])  
        return np.matmul(centered, weights)  
    except ValueError as e:  
        print(f"Failed: {e}")  
        return np.zeros((2, 2))
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numpy + exceptions

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        return np.matmul(centered, weights)
    except ValueError as e:
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```



```
def process_data := fun (data : List (List Rat)) ↦ fun (weights : List (List Rat)) ↦
  ((do
    try
      let mut m := Libraries.numpy.pyNumpyMean data
      let _ <- pyPrintNoop [pyPrintArg s! "Mean: {m}"]
      let mut centered := Libraries.numpy.pyNumpySubtract data [[m, m], [m, m]]
      let mut result := Libraries.numpy.pyNumpyMatmul centered weights
      return result
    catch caught =>
      if (caught).OfKind == "ValueError" then
        let e := caught
        let _ <- pyPrintNoop [pyPrintArg s! "Failed: {e}"]
        let __py_ret_1 := Libraries.numpy.pyNumpyZeros ((2 : Int), (2 : Int))
        return __py_ret_1
      else
        throw caught) :
  PastaLean.PyExceptId _)
```

Design & architecture

The choices that make a *faithful, provable* translation possible — press ↓.

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- **Problem → What we did → Why**, one architectural decision at a time.
- Pure architecture — the elaborator and instance resolution do the heavy lifting.

1 • The logic subset

Not all of Python — only the part that *has* a static meaning.

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```
# in scope — a static, provable core
xs: list[int] = [1, 2, 3]
total = sum(x * x for x in xs)

# out of scope → `pyUnsupported`, flagged
async def fetch(): await conn.read()    # async
result = eval(user_input)                # reflection
def gen(): yield 1                        # laziness
```

2 · One name, many types

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```
-- one Lean name; instance resolution picks the impl per type:
```

```
#check @pyLen
```

```
@pyLen : {a : Type} → [PyLen a] → a → ℤ
```

```
example : Int := pyLen [1, 2, 3]           -- List
```

```
example : Int := pyLen "hello"             -- String
```

```
example : Int := pyLen (Std.HashMap.ofList [(1, 2)]) -- dict
```

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```
#check @PyHAdd
```

```
-- the result type is COMPUTED from the operands (the outParam γ):
```

```
PyHAdd : Type → Type → outParam Type → Type
```

```
example : Int      := (1 : Int) +p (2 : Int)
```

```
example : Float    := (1 : Int) +p (2.0 : Float)
```

```
example : String   := "a" +p "b"
```

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```
-- `self.n = self.n + 1` ↦ record update, rebuilt & returned:
structure Counter where
  n : Int
  deriving Inhabited, Repr, BEq

def Counter.new : Int → Counter := fun (start : Int) ↦
  ({ n := start } : Counter)

def Counter.bump := fun (self : Counter) ↦
  Id.run
  (do
    let mut self := self
    self := { self with n := self.n +p (1 : Int) }
    return self)
```

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- **Why** — exact rationals are decidable and *provable*; you opt into `Float` only when raw speed beats proof.

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- **Problem** — Python `float` is IEEE-lossy; `0.1 + 0.2 ≠ 0.3`, so a proof can't rely on it.
- **We do** — `float` defaults to exact $\mathbb{Q}$ (provable); `--approx` → Lean `Float`; transcendentals → noncomputable $\mathbb{R}$.
- **Why** — exact rationals are decidable and *provable*; you opt into `Float` only when raw speed beats proof.

```
-- Python floats default to EXACT rationals (ℚ):  
example : (1 / 10 + 2 / 10 : Rat) = 3 / 10 := by  
  native_decide  
  
-- --approx picks Lean Float; transcendentals → ℝ:  
#check (Float.sqrt 2.0)  
| Float.sqrt 2.0 : Float  
#check (Real.sqrt 2)  
| √2 : ℝ
```

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New syntax, new library, new container — each is a *local, additive* change.

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```
-- every extension point produces a real runtime name:
#check @[pyLen] -- a container protocol
@[pyLen] : {α : Type} → [PyLen α] → α → ℤ

#check @[pyAppend] -- a list method
@[pyAppend] : {α : Type u_1} → List α → α → List α

#check @[Libraries.numpy.pyNumpyMean] -- a library member
@[pyNumpyMean] : {α : Type} → [PyNumpyScalar α] → List (List α) → Float
```

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When we can't translate faithfully, we refuse *loudly* — never miscompile.

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```
def transform_and_cube := fun a ↦ fun b ↦  
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def total (xs : List Int) : Int :=  
  Id.run do  
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A preference to Non-monadic pure functional code is given while converting. However it will be lowered to Monadic code with `do` , when necessary.

Why lower into a monad at all?

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Tradeoff : Extensibility vs Harder to Prove

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Before • Python + contracts

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def factorial(n: int) → int:
    Requires(n ≥ 0)
    Ensures(Result() ≥ 1)
    result, i = 1, 1
    while i ≤ n:
        Invariant(i ≥ 1)
        Invariant(n - i + 1 ≥ 0)
        Invariant(result ≥ 1)
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        result = result * i
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```

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After · transpiled Lean

```
def factorial := fun (n : Int) ↦
  (do
    let mut result := (1 : Int)
    for i in (PastaLean.pyRange (n +p (1 : Int)) (1 : Int)) do
      let _ := Libraries.passta.pyPassInvariant (decide (i ≥ (1 : Int)))
      let _ := Libraries.passta.pyPassInvariant (decide (n -p i +p 1 ≥ 0))
      let _ := Libraries.passta.pyPassInvariant (decide (result ≥ 1))
      let _ := Libraries.passta.pyPassDecreases (n -p i +p (1 : Int))
      result := result *p i
    return result : Id _)
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A Python `assert` on a pure function becomes `theorem ... := by taste?` — the *Pastafolio* engine searches (`simp` · `grind` · `ring` · `nlinarith` · `fun_induction`) and *splices the found proof back*:

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```
# showcase/linalg/matrix_model.py
def det(a, b, c, d):
    return a*d - b*c

# scaling a 2x2 by k scales det by k²
assert det(k*a, k*b, k*c, k*d) \
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theorem factorial_spec : ⌈ $n \geq (0 : \text{Int})$ ⌋ factorial n ⌋⌋ result ⇒ ⌈ $result \geq (1 : \text{Int})$ ⌋ :=
  by
  mvcgen [factorial, PastaLean.pyRange_forIn, PastaLean.pyRange_forIn_start] invariants
  · ⌋{cur, result} ⇒
    ⌈let i := (cur.prefix.length : Int);
      ( $i \geq (1 : \text{Int}) \wedge n -_p i +_p (1 : \text{Int}) \geq (0 : \text{Int})$ ) ∧ result ≥ (1 : Int)⌋
  simp_all (config := { zetaDelta := true }) [taste_ingr];
  sorry; sorry; omega
```

Proving monadic defs: `mvcgen`

`do`-block functions use `Std.Do`'s `mvcgen` with Hoare triples $\{P\} \text{ prog } \{Q\}$; Python contracts ride along as `Invariant` /

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```
def sum_upto_n(n: int) → int
  Requires(n ≥ 0)
  total = 0
  for i in range(n + 1):
    Invariant(2*total ==
              total + i)
  Ensures(2*total == n*(n + 1))
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theorem sum_upto_n_spec :  $\{n \geq (0 : \text{Int})\} \text{ sum\_to\_n } n \{ \_ \} \Rightarrow \text{True}$  :=
  by
    mvcgen [sum_to_n, PastaLean.pyRange_forIn, PastaLean.pyRange_forIn_start] invariants
    ·  $\downarrow \{cur, total\} \Rightarrow$ 
       $\{let\ i := (cur.prefix.length : \text{Int});$ 
         $(2 : \text{Int}) *_{\text{p}} total = i *_{\text{p}} (i -_{\text{p}} (1 : \text{Int}))\}$ 
      simp_all [taste_ingr]; grind +locals +suggestions
```

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       (2 : Int) * total = i * (i - 1 : Int)}
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`Invariant` / `Ensures` → `pyPassInvariant` / `pyPassEnsures`; `Mvcgen` using Hoare Triples, discharges Goals to prove.

Recap

- *Deterministic transpiler*, not an LLM translator — the elaborator is the oracle.
- Broad language surface: control flow, comprehensions, exceptions, *classes*, libraries.
- *Monadic vs non-monadic* picks the proof strategy:
 - Non-monadic pure function → `taste?`
 - Monadic functions in `do` blocks → `mvcgen`
- *LLMs propose contracts*; Lean checks them.

Pasta + Python + Lean = Python you can *prove*.

PastaLean

Thank you

Questions?

PA²TaV