Installation

pip install -U taichi # Install / upgrade ti gallery # Run demo gallery ti example # More examples

Ouick start

```
import taichi as ti
ti.init(arch=ti.cpu)
# Or use another backend
# [ti.cuda, ti.vulkan, ti.opengl, ti.metal]
# Args and return of a kernel must be type hinted
def monte_carlo_pi(n: int) -> float:
    total = 0
    for i in range(n): # A parallel for loop
       x = ti.random()
       y = ti.random()
        if x*x + y*y < 1:
            total += 1
    return 4 * total / n
print(monte_carlo_pi(100000))
```

Kernels and functions

@ti.kernel: Entrance for Taichi's JIT to take control. Must be called from Python scope. Require type hints for the arguments and the return value. Can return at most one scalar or vector or matrix. Top-level for loops are automatically parallelized.

Oti.func: Must be called by kernels or other Taichi functions. Recommend type hints for arguments and return values. Can return multiple values of scalars, vectors, matrices, and structs.

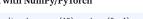
Top-level for loops are automatically parallelized

```
x = ti.field(dtype=int, shape=100)
Oti kernel
def loop_field():
    for i in range(100): # A parallelized loop
       for j in range(10): # Not parallelized
    for i in x: # Also a parallelized loop
       x[i] = i
```

Serialize a top-level for loop

```
@ti.kernel
def config_loop():
   # Serialize the *next* for loop
   ti.loop_config(serialize=True)
   for i in range(100): # Serial
   for i in range(100): # Parallel
```

Interact with NumPy/PyTorch



x = numpy/torch.arange(12).reshape(3, 4) def fill_array(arr: ti.types.ndarray()): print(arr.shape, arr.dtype) for i, i in arr: arr[i, j] = i + j print(arr[i, j]) fill_array(x) # zero copy if on the same device

```
Primitive types (int, unsigned int and float in C)
                  u8 # uchar
                                    f16 # half
i8 # char
i16 # short
                  u16 # ushort
                                    f32 # float
i32 # int
                  u32 # uint
                                    f64 # double
                  u64 # ulong
```

Vector and matrix types

m = mat2(1, 2, 3, 4)

Data types

i64 # long

• Vector: ti.types.vector(dim, dtype)

```
vec3 = ti.types.vector(3, float)
v = vec3(1, 2, 3)
• Matrix: ti.types.matrix(n, m, dtype)
mat2 = ti.tvpes.matrix(2, 2, float)
```

Struct types: ti.types.struct(**kwargs)

```
sphere = ti.types.struct(
   center=vec3, radius=float
s = sphere(vec3(1, 2, 3), 1.0)
```

Ndarray types:

```
ti.tvpes.ndarrav(
   dtype, # Data type of the array.
   ndim # Number of field dimentions
```

```
img2d_type = ti.types.ndarray(ndim=3)
noise_img = np.random.random((400, 400, 3))
Oti kernel
def process(img: img2d_type):
    for I in ti.grouped(img):
       r, g, b = img[I]
process(noise_img)
```

Typecasting

```
x = ti.f32(1)
y = ti.u8(x) # Equivalent to x.cast(ti.u8)
z = float(y)
u = vec3(0.5, 1.0, 1.5)
v = u.cast(int) # Cast each entry to int
```

Performance tuning



Kernel profiler (CPU and CUDA only)

Analyze the performance of Taichi kernels

```
ti.init(arch=ti.cpu, kernel_profiler=True)
@ti.kernel
def some_kernel():
some kernel()
ti.profiler.print_kernel_profiler_info()
```

Configure loops

Set arguments in ti.loop_config() to control the next for loop:

- 1. Set the number of threads in a block on GPU: block dim=8.
- 2. Set the number of threads to use on CPU: parallelize=8.
- 3. Whether to let the for loop execute serially: serialize=True.

```
Data containers
```

Scalar field: field(dtype, ...)

```
f = ti.field(int, shape=(3, 3, 3))
f[0, 1, 2] = 3 # Indexing with three integers
x = ti.field(int, shape=()) # 0-D field
x[None] = 1.0 # Use None to index 0-D field
```

Vector field: Vector.field(dim, dtype, ...)

```
f = ti.Vector.field(3, float, (10, 10))
f[0, 1] = 1, 2, 3 # Each entry is a 3D vector
```

Matrix field: Matrix.field(n, m, dtype, ...)

```
f = ti.Matrix.field(2, 2, float, shape=(10, 10))
f[0, 1] = mat2(1) # Each entry is a 2x2 matrix
```

Struct field: obj.field(shape)

```
sphere = ti.types.struct(center=vec3, radius=
f = sphere.field(shape=100)
```

Fill a field with a scalar

f.fill(1)

Copy data from/to NumPy array/PyTorch tensor

```
f.from_numpy(arr)
arr = f.to_numpy()
f.from_torch(tensor)
tensor = f.to_torch()
```

Loop over a field in parallel

```
f = ti.field(int, shape=(100, 100, 100))
Oti kernel
def loop_field():
    for i, j, k in ti.ndrange(100, 100, 100):
        f[i, j, k] = i + j + k
    # Equivalent to the above
    for i, j, k in f:
        f[i, j,k] = i + j + k
    # Equivalent to the above
    for I in ti.grouped(f):
        # I = [i, j, k] is a 3D int vector
        f[I] = I[0] + I[1] + I[2]
```

Switch data layout between AOS and SOA

```
u = ti.Vector.field(
   3, float, shape=100, layout=ti.Layout.AOS)
# array of structs [x0,y0,z0,x1,y1,z1...,]
v = ti.Vector.field(
   3, float, shape=100, layout=ti.Layout.SOA)
# struct of arrays [x0,x1...,,y0,y1...,,z0,z1...,]
```

Math functions

import taichi as ti	import taichi.math as tm
ti.cos(x)	tm.cross(u,v)
ti.sin(x)	tm.dot(u,v)
ti.acos(x)	tm.fract(x)
ti.asin(x)	tm.mod(x,y)
ti.atan2(y, x)	tm.normalize(v)
ti.exp(x)	tm.smoothstep(e0,e1,x)
ti.log(x)	tm.mix(x,y,a)
ti.ceil(x, dtype)	tm.step(edge,x)
ti.floor(x, dtype)	tm.degrees(x)
ti.round(x, dtype)	tm.radians(x)
ti.pow(x, a)	tm.clamp(x,xmin,xmax)
ti.tan(x)	tm.length(v)
ti.tanh(x)	tm.log2(x)
ti.sqrt(x)	tm.inverse(mat)
ti.max(x,y,)	tm.isnan(x)
ti.min(x,y,)	tm.isinf(x)
ti.random(dtype)	tm.sign(x)

Arithmetic operators

Operators

```
-x, x + y, x - y, x * y,
x / y, # returns a floating point. 5 / 2 = 2.5
x // y # floor of x / y. 5.0 / 2.0 = 2.0
x % y # remainder of x / y. x & y can be floats
x ** y # x to the power of y
A @ B # matrix multiplication
```

Comparison operators

```
x == y, x != y, x > y, x < y, x >= y, x <= y
```

Logical operators

```
not, or, and
```

Bitwise operators

```
~x, x & y, x ^ y, x | y, x << y, x >> y
```

Data-oriented programming



Data-oriented class

When you have data maintained in the Python scope (such as time or user input events) and you want the kernels to track their changes, you can organize them into a data-oriented class.

```
@ti.data oriented
class TiArray:
    def __init__(self, n):
        self.x = ti.field(dtype=ti.i32, shape=n)
    @ti.kernel
    def inc(self):
        for i in self.x:
            self.x[i] += 1
a = TiArray(32)
a.inc()
```

dataclass

A dataclass is a wrapper of ti.types.struct. You can define Taichi functions as its methods and call these methods in the Taichi scope.

```
@ti.dataclass
class Sphere:
   center: vec3
   radius: float
   def area(self): # A Taichi function as method
       return 4 * pi * self.radius**2
   sphere = Sphere(vec3(0), radius=1.0)
   print(sphere.area())
```

Global settings



You can config Taichi by passing arguments to the ti.init() call:

- 1. Choose a backend: arch=ti.cuda.
- 2. Enable debug mode: debug=True.
- Enable dynamic index: dynamic_index=True.
- 4. Set floating precision: default_fp=ti.f64.
- 5. Set integer precision: default_ip=ti.i64. 6. Set random seed: random seed=0.
- 7. Disable offline cache: offline cache=False.
- 8. Enable packed mode: packed=True.

device_memory_GB=1.3.

9. Set logging level: log_level=ti.ERROR. 10. Set pre-allocated memory size for CUDA:

Visualization

GUI system

```
pixels = ti.Vector.field(3, float, (640, 480))
gui = ti.GUI('Window Title', res=(640, 480))
while gui.running:
    gui.set_image(pixels)
    gui.show()
```

GGUI system

```
pixels = ti.Vector.field(3, float, (640, 480))
window = ti.ui.Window('Window Title', (640, 480))
canvas = window.get_canvas()
while window.running:
   canvas.set_image(pixels)
   window.show()
window.save_image(filename) # save image file
```

2D canvas drawing API

```
canvas.set background color(color)
canvas.triangles(vertices, color, indices,
      per_vertex_color)
canvas.circles(vertices, radius, color,
      per_vertex_color)
canvas.lines(vertices, width, indices, color,
      per_vertex_color)
```

3D scene drawing API

```
scene.lines(vertices, width, indices, color,
      per_vertex_color)
scene.mesh(vertices, indices, normals, color,
      per_vertex_color)
scene.particles(vertices, radius, color,
      per_vertex_color)
```

Debugging ti.init(..., debug=True)

Debug mode can help check access out-of-bound errors and allow you to assert in Taichi kernels on CPU and CUDA backends.

```
ti.init(arch=ti.cpu, debug=True)
f = ti.field(int, shape=(5, 5))
print(f[7, 7]) # Only raise error in debug mode!
```

Runtime assert in Taichi kernels and functions

```
x = ti.field(dtype=ti.f32, shape=128)
Qti.kernel
def foo():
   for i in x:
        assert x[i] >= 0
        x[i] = ti.sqrt(x[i])
```

Runtime print in Taichi kernels and functions

```
Qti.kernel
def inside_taichi_scope(x: float):
    # print is supported on cpu, cuda
    # and vulkan backends
    print('hello', x) # Cannot use f-string
```

Compile-time static-print

```
x = ti.field(ti.f32, (2, 3))
def print_field_attributes():
    ti.static_print(x.shape, x.dtype)
```

Compile-time static-assert

```
@ti.func # Assuming dst and src are fields
def copy(dst: ti.template(), src: ti.template()):
    ti.static_assert(dst.shape == src.shape)
    for I in ti.grouped(src):
        dst[I] = src[I]
```