# Data-driven Computer Animation

**Tutorial 5 – Introduction to Taichi** 

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#### What is Taichi

A programming language designed for computer graphics

- Productivity
  - Friendly learning curve
  - Shorter code, higher perf.
- Portability
  - Multi-backend support
- Performance
  - Optimized for bandwidth, locality and load balancing

# Why Taichi

- Productivity
- Portability
- Performance

```
constraint.cpp

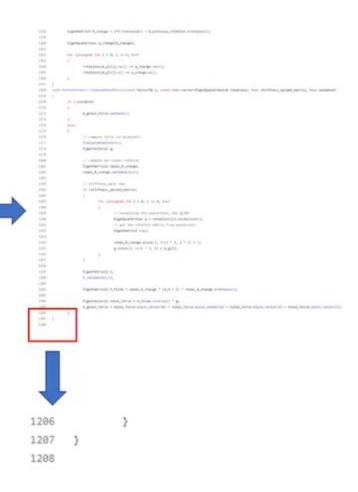
constraint.h

constraint_attachment.cpp

constraint_penalty.cpp

constraint_spring.cpp

constraint_tet.cpp
```

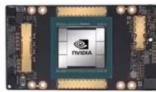


# Why Taichi

- Productivity
- Portability
- Performance















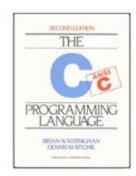


# Why Taichi

- Productivity
- Portability
- Performance







# ti.init()

Entrance to all Taichi projects

```
import taichi as titi.init(arch=ti.gpu)
```

# ti.init()

- arch:
  - ti.cpu/ti.gpu/ti.arm/ti.cuda...
- import taichi as ti
- ti.init(arch=ti.gpu)

	CPU	CUDA	OpenGL	Apple Metal	Vulkan
Windows	YES	YES	YES	NO	WIP
Linux	YES	YES	YES	NO	YES
macOS	YES	NO	NO	YES	WIP

# ti.init()

- To specify the GPU ID
  - for CUDA: export CUDA\_VISIBLE\_DEVICES=[gpuid]
  - for Vulkan: export TI\_VISIBLE\_DEVICES=[gpuid]

	CPU	CUDA	OpenGL	Apple Metal	Vulkan
Windows	YES	YES	YES	NO	WIP
Linux	YES	YES	YES	NO	YES
macOS	YES	NO	NO	YES	WIP

# Python vs Taichi

```
import taichi as ti
ti.init(arch=ti.cpu)
d = 1
                                                       Python-scope
def foo():
   d python = d
   print("d_python =", d_python)
@ti.kernel
def bar():
                                                       Taichi-scope
   d taichi = d
   print("d taichi =", d taichi)
d = d + 1 # d = 2
foo() # d python = 2
bar() # d_taichi = 2
                                                     Only codes in ti.kernel and ti.func are
d = d + 1 # d = 3
                                                     in Taichi-scope
foo() # d python = 3
bar() # d_taichi = 2
```

# Tachi – Data types

- signed integers: ti.i8, ti.i16, ti.i32, ti.i64
- unsigned integers: ti.u8, ti.u16, ti.u32, ti.u64
- floating points: ti.f32, ti.f64

Backend	<b>i8</b>	<b>i16</b>	<b>i32</b>	<b>i64</b>	u8	<b>u16</b>	u32	u64	f16	f32	f64
CPU	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>
CUDA	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>	<b>~</b>
OpenGL	×	×	<b>~</b>	0	×	×	×	×	×	<b>~</b>	<b>~</b>
Metal	<b>~</b>	<b>~</b>	<b>~</b>	×	<b>~</b>	<b>✓</b>	<b>~</b>	×	×	<b>~</b>	×
Vulkan	0	0	<b>~</b>	0	0	0	<b>~</b>	0	<b>~</b>	<b>~</b>	0

O: Requiring extensions for the backend.

## Data Type Aliases

 Default types can be changed using the configuration option default\_ip and default\_fp

```
ti.init(default_ip=ti.i64, default_fp=ti.f64)

@ti.kernel
def example_cast() -> int: # the returned type is ti.i64
    x = 3.14  # x is of ti.f64 type
    y = int(x) # equivalent to ti.i64(x)
    return y
```

# Data Type Aliases

Do not mix up Taichi's int and other int

```
x = numpy.array([1, 2, 3, 4], dtype=int) # NumPy's int64 type
y = int(3.14) # Python's built-in int type
```

## Type Casts

- Implicit casts
  - static types within the Taichi scope

```
import taichi as ti

ti.init(arch=ti.cpu)

def foo():
    a = 1
    a = 2.7
    print(a)

foo() #2.7
```

```
import taichi as ti

ti.init(arch=ti.cpu)

@ti.kernel
def foo():
    a = 1
    a = 2.7
    print(a)

foo() #2
```

## Type Casts

variable = ti.cast(variable, type)

```
import taichi as ti
ti.init(arch=ti.cpu)
@ti.kernel
def foo():
    a = 1.7
    b = ti.cast(a, ti.i32)
    c = ti.cast(b, ti.f32)
    print("b =", b) # b = 1
    print("c =", c) # c = 1.0
foo()
```

## Compound Types

- Using ti.types to create compound types including:
  - vector / matrix / struct

```
import taichi as ti
ti.init(arch=ti.cpu)
vec3f = ti.types.vector(3, ti.f32)
mat2f = ti.types.matrix(2, 2, ti.f32)
ray = ti.types.struct(ro=vec3f, rd=vec3f, l=ti.f32)
@ti.kernel
def foo():
   a = vec3f(0.0)
   print(a)
                             # [0.0, 0.0, 0.0]
   d = vec3f(0.0, 1.0, 0.0)
   print(d)
                            # [0.0, 1.0, 0.0]
   B = mat2f([[1.5, 1.4], [1.3, 1.2]])
   print("B =", B) # B = [[1.5, 1.4], [1.3, 1.2]]
   r = ray(ro=a, rd=d, l=1)
   print("r.ro =", r.ro) # r.ro = [0.0, 0.0, 0.0]
   print("r.rd =", r.rd) # r.rd = [0.0, 1.0, 0.0]
foo()
```

## Compound Types

- Predefined keywords for compound types:
  - ti.Vector / ti.Matrix / ti.Struct

```
import taichi as ti
ti.init(arch=ti.cpu)
@ti.kernel
def foo():
   a = ti.Vector([0.0, 0.0, 0.0])
   print(a)
                             # [0.0, 0.0, 0.0]
   d = ti.Vector([0.0, 1.0, 0.0])
   print(d)
              # [0.0, 1.0, 0.0]
   B = ti.Matrix([[1.5, 1.4], [1.3, 1.2]])
   print("B =", B) # B = [[1.5, 1.4], [1.3, 1.2]]
   r = ti.Struct(v1=a, v2=d, l=1)
   print("r.v1 =", r.v1) # r.v1 = [0.0, 0.0, 0.0]
   print("r.v2 =", r.v2) # r.v2 = [0.0, 1.0, 0.0]
foo()
```

## Compound Types - Indexing

Access compound elements using [i,j,k,...] indexing

```
import taichi as ti

ti.init(arch=ti.cpu)

@ti.kernel
def foo():
    a = ti.Vector([1.0, 2.0, 3.0])
    print(a[1]) # 2.0

    B = ti.Matrix([[1.5, 1.4], [1.3, 1.2]])
    print(B[1,0]) # 1.3

foo()
```

"a global N-d array of elements"

```
heat_field = ti.field(dtype=ti.f32, shape=(256, 256))
```

- "a global N-d array of elements"
  - global: can be read/written from both the Taichi-scope and the Python-scope
  - N-d: (Scalar: N=0), (Vector: N=1), (Matrix: N=2), (N = 3, 4, 5, ...)
  - elements: scalar, vector, matrix, struct

- "a global N-d array of elements"
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  - N-d: (Scalar: N=0), (Vector: N=1), (Matrix: N=2), (N = 3, 4, 5, ...)
  - elements: scalar, vector, matrix, struct

• "3D gravitational field in a 256x256x128 room"

```
gravitational_field = ti.Vector.field(n = 3,dtype=ti.f32,shape=(256,256,128))
```

"2D strain-tensor field in a 64x64 grid"

```
strain_tensor_field = ti.Matrix.field(n = 2,m = 2, dtype=ti.f32, shape=(64,64))
```

• "a global scalar that I want to access in a Taichi kernel"

```
global_scalar = ti.field(dtype=ti.f32, shape=())
```

- "a global N-d array of elements"
  - global: can be read/written from both the Taichi-scope and the Python-scope
  - N-d: (Scalar: N=0), (Vector: N=1), (Matrix: N=2), (N = 3, 4, 5, ...)
  - elements: scalar, vector, matrix, struct
  - access elements in a field using [i,j,k,...] indexing

```
import taichi as ti

ti.init(arch=ti.cpu)

pixels = ti.field(dtype=float, shape=(16, 8))

pixels[1, 2] = 42.0
```

```
import taichi as ti

ti.init(arch=ti.cpu)

vf = ti.Vector.field(3, ti.f32, shape=4)

@ti.kernel
def foo():
    v = ti.Vector([1, 2, 3])
    vf[0] = v
```

- "a global N-d array of elements"
  - global: can be read/written from both the Taichi-scope and the Python-scope
  - N-d: (Scalar: N=0), (Vector: N=1), (Matrix: N=2), (N = 3, 4, 5, ...)
  - elements: scalar, vector, matrix, struct
  - access elements in a field using [i,j,k,...] indexing
    - Special case, access a zero-d field using [None]

```
zero_d_scalar = ti.field(ti.f32, shape=())
zero_d_scalar[None] = 1.5

zero_d_vec = ti.Vector.field(2, ti.f32, shape=())
zero_d_vec[None] = ti.Vector([2.5, 2.6])
```

- "a global N-d array of elements"
  - global: can be read/written from both the Taichi-scope and the Python-scope
  - N-d: (Scalar: N=0), (Vector: N=1), (Matrix: N=2), (N = 3, 4, 5, ...)
  - elements: scalar, vector, matrix, struct
  - access elements in a field using [i,j,k,...] indexing
    - Special case, access a zero-d field using [None]

```
zero_d_scalar = ti.field(ti.f32, shape=())
zero_d_scalar[None] = 1.5

zero_d_vec = ti.Vector.field(2, ti.f32, shape=())
zero_d_vec[None] = ti.Vector([2.5, 2.6])
```

# ti.grouped()

- Taichi provides ti.grouped syntax which supports grouping loop indices into a ti.Vector.
- It enables dimensionality-independent programming, i.e., code are adaptive to scenarios of different dimensionalities automatically

```
# without ti.grouped
for I in ti.ndrange(2, 3):
    print(I)
prints 0, 1, 2, 3, 4, 5
```

```
# with ti.grouped
for I in ti.grouped(ndrange(2, 3)):
    print(I)
prints [0, 0], [0, 1], [0, 2], [1, 0], [1, 1], [1, 2]
```

## ti.grouped()

- Taichi provides ti.grouped syntax which supports grouping loop indices into a ti.Vector.
- It enables dimensionality-independent programming, i.e., code are adaptive to scenarios of different dimensionalities automatically

```
import taichi as ti
ti.init()

a = ti.Matrix.field(n=2, m=3, dtype=ti.f32, shape=(2, 2))
@ti.kernel
def test():
    for i in ti.grouped(a):
        # a[i] is a 2x3 matrix
        a[i] = [[1, 1, 1], [1, 1, 1]]
```

#### Matrix Size Consideration

• Matrix operations are unrolled at compile time. For performance reasons, it is recommended that you keep your matrices small.

```
import taichi as ti
ti.init()
a = ti.Matrix.field(n=2, m=3, dtype=ti.f32, shape=(2, 2))
@ti.kernel
def test():
    for i in ti.grouped(a):
        # a[i] is a 2x3 matrix
        a[i] = [[1, 1, 1], [1, 1, 1]]
        # The assignment is unrolled to the following at compile time:
        \# a[i][0, 0] = 1
        # a[i][0, 1] = 1
        \# a[i][0, 2] = 1
        \# a[i][1, 0] = 1
        \# a[i][1, 1] = 1
        # a[i][1, 2] = 1
```

#### Matrix Size Consideration

• Matrix operations are unrolled at compile time. For performance reasons, it is recommended that you keep your matrices small.

```
import taichi as ti
ti.init()
a = ti.Matrix.field(n=2, m=3, dtype=ti.f32, shape=(2, 2))
@ti.kernel
def test():
    for i in ti.grouped(a):
        # a[i] is a 2x3 matrix
        a[i] = [[1, 1, 1], [1, 1, 1]]
        # The assignment is unrolled to the following at compile time:
        \# a[i][0, 0] = 1
        # a[i][0, 1] = 1
        \# a[i][0, 2] = 1
        \# a[i][1, 0] = 1
        \# a[i][1, 1] = 1
        \# a[i][1, 2] = 1
```

#### Matrix Size Consideration

 Workaround: When declaring a matrix field, leave large dimensions to the fields, rather than to the matrices. If you have a 3x2 field of 64x32 matrices:

- Not recommended: ti.Matrix.field(64, 32, dtype=ti.f32, shape=(3, 2))
- Recommended: ti.Matrix.field(3, 2, dtype=ti.f32, shape=(64, 32))

## Computation Kernel

- A Python function decorated by @ti.kernel is a Taichi kernel
  - Taichi kernels can only be called from the Python scope

```
import taichi as ti
ti.init(arch=ti.cpu)
@ti.kernel
def foo():
    print("foo")
@ti.kernel
def bar():
    print("bar")
foo()
bar()
```

```
import taichi as ti
ti.init(arch=ti.cpu)
def foo():
    print("foo")
    bar()
@ti.kernel
def bar():
    print("bar")
foo()
```

```
import taichi as ti
ti.init(arch=ti.cpu)
@ti.kernel
def foo():
    print("foo")
    bar()
@ti.kernel
def bar():
    print("bar")
foo()
```

 For loops at the outermost scope in a Taichi kernel is automatically parallelized

```
@ti.kernel
def fill():
    for i in range(10): # Parallelized
        x[i] += i
        s = 0
        for j in range(5): # Serialized in each parallel thread
            s += j
        y[i] = s
    for k in rang(20): # Parallelized
        z[k] = k
```

Outermost scope ?

```
import taichi as ti
ti.init(arch=ti.cpu)
@ti.kernel
def foo(k: ti.i32):
    for i in range(10): # Parallelized :-)
        if k > 42:
            . . .
@ti.kernel
def bar(k: ti.i32):
    if k > 42:
        for i in range(10): # Serial :-(
```

Design your for loops for best performance

```
def my_for_loop():
    for i in range(10): # I don't want to parallelize this for
        for j in range(100): # I want to parallelize this for
            . . .
my_for_loop()
def my_for_loop():
    for i in range(10):
        my_taichi_for()
```





break is NOT supported in the parallel for-loops

```
@ti.kernel
def foo():
  for i in range(10):
      break # Error!
@ti.kernel
def foo():
  for i in range(10):
      for j in range(10):
          break # OK!
```

- Race condition
  - Taichi uses += as an atomic add
  - The compiler optimizes for unnecessary atomic operations

```
@ti.kernel
def sum():
    for i in range(10):
        # 1. OK
        total[None] += x[i]

    # 2. OK
        ti.atomic_add(total[None], x[i])

    # 3. data race
        total[None] = total[None] + x[i]
```

- Types of for-loops in Taichi
  - range-for: loops over a range, identical to Python range-for
  - struct-for: loops over a ti.field, only lives at the outermost scope

```
import taichi as ti

ti.init(arch=ti.cpu)

N = 10
x = ti.field(dtype=ti.i32, shape=N)

@ti.kernel
def foo():
    for i in range(N):
        x[i] = i

foo()
```

```
import taichi as ti

ti.init(arch=ti.cpu)

N = 10
x = ti.Vector.field(2,dtype=ti.i32, shape=(N,N))

@ti.kernel
def foo():
    for i,j in x:
        x[i,j] = ti.Vector([i, j])

foo()
```

- Types of for-loops in Taichi
  - range-for: loops over a range, identical to Python range-for
  - struct-for: loops over a ti.field, only lives at the outermost scope

```
import taichi as ti

ti.init(arch=ti.cpu)

N = 10
x = ti.field(dtype=ti.i32, shape=N)

@ti.kernel
def foo():
    for i in range(N):
        x[i] = i

foo()
```

```
import taichi as ti

ti.init(arch=ti.cpu)

N = 10
x = ti.Vector.field(2,dtype=ti.i32, shape=(N,N))

@ti.kernel
def foo():
    for i,j in x:
        x[i,j] = ti.Vector([i, j])

foo()
```

A kernel can accept multiple arguments.

However, it's important to note that you can't pass arbitrary Python objects to a kernel.

Python objects can be dynamic and may contain data that the Taichi compiler cannot recognize.

- Scalars
- ti.types.matrix()
- ti.types.vector()
- ti.types.struct()
- ti.types.ndarray()
- •

- Passed by value
  - ti.types.matrix()
  - ti.types.vector()
  - ti.types.struct()
- Passed by reference
  - ti.types.ndarray()
  - ti.template()

Must have type hint

```
transform_type = ti.types.struct(R=ti.math.mat3, T=ti.math.vec3)
pos_type = ti.types.struct(x=ti.math.vec3, trans=transform_type)

@ti.kernel
def kernel_with_nested_struct_arg(p: pos_type) -> ti.math.vec3:
    return p.trans.R @ p.x + p.trans.T

trans = transform_type(ti.math.mat3(1), [1, 1, 1])
p = pos_type(x=[1, 1, 1], trans=trans)
print(kernel_with_nested_struct_arg(p)) # [4., 4., 4.]
```

#### Kernel Return Value

- Must have the type hint
- Could either be a scalar, ti.types.matrix(), or ti.types.vector()
- In CPU and CUDA backend, could also be ti.types.struct()
- If the return value is a vector or matrix, please ensure that it contains no more than 32 elements. (Warning otherwise)

#### Kernel Return Value

At most one return value

```
vec2 = ti.math.vec2

@ti.kernel
def test(x: float, y: float) -> vec2: # Return value must be type hinted
     # Return x, y # Compilation error: Only one return value is allowed
    return vec2(x, y) # Fine
```

At most one return statement

```
@ti.kernel
def test_sign(x: float) -> float:
    if x >= 0:
        return 1.0
    else:
        return -1.0
# Error: multiple return statements
```

#### Taichi Function

 Taichi functions are fundamental units of a kernel and can only be called from within a kernel or another Taichi function.

## Kernel vs Function

	Kernel	Taichi Function		
Call scope	Python scope	Taichi scope		
Type hint arguments	Mandatory	Recommended		
Type hint return values	Mandatory	Recommended		
Return type	<ul> <li>Scalar</li> <li>ti.types.matrix()</li> <li>ti.types.vector()</li> <li>ti.types.struct() (Only on LLVM-based backends)</li> </ul>	<ul><li>Scalar</li><li>ti.types.matrix()</li><li>ti.types.vector()</li><li>ti.types.struct()</li><li></li></ul>		
Maximum number of elements in arguments	<ul><li>32 (OpenGL)</li><li>64 (otherwise)</li></ul>	Unlimited		
Maximum number of return values in a return statement	1	Unlimited		