Slice 层

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初始示例代码

```
import numpy as np
from cuda import cudart
import tensorrt as trt
nIn, cIn, hIn, wIn = 1, 3, 4, 5 # 输入张量 NCHW
data = np.arange(cIn, dtype=np.float32).reshape(cIn, 1, 1) * 100 + <math>np.arange(hIn).reshape(1, hIn, 1) *
10 + np.arange(wIn).reshape(1, 1, wIn) # 输入数据
data = data.reshape(nIn, cIn, hIn, wIn).astype(np.float32)
np.set_printoptions(precision=8, linewidth=200, suppress=True)
cudart.cudaDeviceSynchronize()
logger = trt.Logger(trt.Logger.ERROR)
builder = trt.Builder(logger)
network = builder.create_network(1 << int(trt.NetworkDefinitionCreationFlag.EXPLICIT_BATCH))</pre>
config = builder.create_builder_config()
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (nIn, cIn, hIn, wIn))
sliceLayer = network.add_slice(inputT0, (0, 0, 0, 0), (1, 2, 3, 4), (1, 1, 1, 1))
network.mark_output(sliceLayer.get_output(0))
engineString = builder.build_serialized_network(network, config)
engine = trt.Runtime(logger).deserialize_cuda_engine(engineString)
context = engine.create_execution_context()
_, stream = cudart.cudaStreamCreate()
inputH0 = np.ascontiguousarray(data.reshape(-1))
outputH0 = np.empty(context.get_binding_shape(1), dtype=trt.nptype(engine.get_binding_dtype(1)))
_, inputD0 = cudart.cudaMallocAsync(inputH0.nbytes, stream)
_, outputD0 = cudart.cudaMallocAsync(outputH0.nbytes, stream)
cudart.cudaMemcpyAsync(inputD0, inputH0.ctypes.data, inputH0.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyHostToDevice, stream)
context.execute_async_v2([int(inputD0), int(outputD0)], stream)
cudart.cudaMemcpyAsync(outputH0.ctypes.data, outputD0, outputH0.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyDeviceToHost, stream)
cudart.cudaStreamSynchronize(stream)
print("inputH0 :", data.shape)
```

```
print(data)
print("outputH0:", outputH0.shape)
print(outputH0)

cudart.cudaStreamDestroy(stream)
cudart.cudaFree(inputD0)
cudart.cudaFree(outputD0)
```

• 输入张量形状 (1,3,4,5), 百位、十位、个位分别表示 CHW 维编号

```
[100. 101. 102. 103. 104.]
                                        200. 201.
                                                   202.
                                                        203.
                                                              204.
          110. 111. 112. 113. 114.
                                        210.
                                             211.
                                                   212.
                                                         213.
                                                              214.
         120. 121. 122. 123. 124.
                                        220.
                                             221.
                                                   222. 223.
                                                              224.
33. 34. \rfloor [130. 131. 132. 133. 134. \rfloor [230.
                                             231. 232. 233.
```

• 输出张量形状 (1,2,3,4), 以 (0,0,0) 元素为起点, 切出 (1,2,3,4) 形状的张量, 各维上的步长为 (1,1,1)

```
\left[ \begin{bmatrix} 0. & 1. & 2. & 3. \\ 10. & 11. & 12. & 13. \\ 20. & 21. & 22. & 23. \end{bmatrix} \begin{bmatrix} 100. & 101. & 102. & 103. \\ 110. & 111. & 112. & 113. \\ 120. & 121. & 122. & 123. \end{bmatrix} \right]^{-1}
```

start & shape & stride

```
sliceLayer = network.add_slice(inputT0, (0, 0, 0, 0), (0, 0, 0, 0), (0, 0, 0, 0))
sliceLayer.start = (0, 0, 0, 0) # 重设起点位置
sliceLayer.shape = (1, 2, 3, 4) # 重设输出张量形状
sliceLayer.stride = (1, 1, 1, 1) # 重设切割步长, 1 为连续提取元素
```

• 输出张量形状 (1,2,3,4), 结果与初始示例代码相同

mode

```
sliceLayer = network.add_slice(inputT0, (0, 0, 0, 0), (1, 2, 3, 4), (1, 2, 2, 2)) sliceLayer.mode = trt.SliceMode.WRAP
```

• 指定 mode=trt.SliceMode.WRAP,输出张量形状 (1,2,3,4),超出边界的元素从起始元素继续切片

$$\left[\left[\begin{bmatrix} 0. & 2. & 4. & 1. \\ 20. & 22. & 24. & 21. \\ 0. & 2. & 4. & 1. \end{bmatrix} \begin{bmatrix} 200. & 202. & 204. & 201. \\ 220. & 222. & 224. & 221. \\ 200. & 202. & 204. & 201. \end{bmatrix} \right] \right]$$

• 指定 mode=trt.SliceMode.CLAMP,输出张量形状 (1,2,3,4),超出边界的元素取边界值

$$\left[\left[\begin{bmatrix} 0. & 2. & 4. & 4. \\ 20. & 22. & 24. & 24. \\ 30. & 32. & 34. & 34. \end{bmatrix} \left[\begin{matrix} 200. & 202. & 204. & 204. \\ 220. & 222. & 224. & 224. \\ 230. & 232. & 234. & 234. \end{bmatrix} \right] \right]$$

- 指定 mode=trt.SliceMode.FILL、输出张量形状 (1,2,3,4)、超出边界的元素取固定值、默认值 0
- 需要设置其他固定值的情况,参见后面的 fill mode + set input 部分

$$\left[
\begin{bmatrix}
0. & 2. & 4. & 0. \\
20. & 22. & 24. & 0. \\
0. & 0. & 0. & 0.
\end{bmatrix}
 \begin{bmatrix}
200. & 202. & 204. & 0. \\
220. & 222. & 224. & 0. \\
0. & 0. & 0. & 0.
\end{bmatrix}
\right]$$

• 指定 mode=trt.SliceMode.REFECT,输出张量形状 (1,2,3,4),超出边界的元素折返取值

```
\left[ \begin{bmatrix} 0. & 2. & 4. & 2. \\ 20. & 22. & 24. & 22. \\ 20. & 22. & 24. & 22. \end{bmatrix} \begin{bmatrix} 200. & 202. & 204. & 0. \\ 220. & 222. & 224. & 0. \\ 220. & 222. & 224. & 0. \end{bmatrix} \right]
```

• 可用的模式

trt.SliceMode 名	说明
DEFAULT	默认模式,超出边界就报错
WRAP	data[i] = data[i%w]
CLAMP	data[i] = data[w-1]
FILL	data[i] = fillValue
REFLECT	data[i] = data[(w-1-i%w)(<i>i/w%2</i>)+(<i>i%w</i>)(1-i/w%2)]

set_input

fill mode + set_input

```
constantLayer = network.add_constant([1], np.array([-1], dtype=np.float32))
sliceLayer = network.add_slice(inputT0, (0, 0, 0, 0), (1, 2, 3, 4), (1, 2, 2, 2))
sliceLayer.mode = trt.SliceMode.FILL
sliceLayer.set_input(4, constantLayer.get_output(0)) # 在 4 号位上设置固定值
```

• 输出张量形状 (1,2,3,4), 超出边界的元素取固定值

$$\left[\begin{bmatrix} 0. & 2. & 4. & -1. \\ 20. & 22. & 24. & -1. \\ -1. & -1. & -1. & -1. \end{bmatrix} \begin{bmatrix} 200. & 202. & 204. & -1. \\ 220. & 222. & 224. & -1. \\ -1. & -1. & -1. \end{bmatrix} \right]$$

静态 set_input

```
constantLayer0 = network.add_constant([4], np.array([0, 0, 0, 0], dtype=np.int32))
constantLayer1 = network.add_constant([4], np.array([1, 2, 3, 4], dtype=np.int32))
constantLayer2 = network.add_constant([4], np.array([1, 1, 1, 1], dtype=np.int32))
sliceLayer = network.add_slice(inputT0, (0, 0, 0, 0), (0, 0, 0, 0), (0, 0, 0, 0))
#sliceLayer.set_input(0,inputT0)
# 0
号输入是被 slice 的张量
sliceLayer.set_input(1, constantLayer0.get_output(0)) # 2、3、4 号输入分别对应起点、形状和步长
sliceLayer.set_input(2, constantLayer1.get_output(0))
sliceLayer.set_input(3, constantLayer2.get_output(0))
```

• 输出张量形状 (1,2,3,4), 结果与初始示例代码相同

动态 set_input

```
import numpy as np
from cuda import cudart
import tensorrt as trt

nIn, cIn, hIn, wIn = 1, 3, 4, 5
data0 = np.arange(cIn, dtype=np.float32).reshape(cIn, 1, 1) * 100 + np.arange(hIn).reshape(1, hIn, 1) *
10 + np.arange(wIn).reshape(1, 1, wIn)
data0 = data0.reshape(nIn, cIn, hIn, wIn).astype(np.float32)
```

```
data1 = np.array([0, 0, 0, 0], dtype=np.int32)
data2 = np.array([1, 2, 3, 4], dtype=np.int32)
data3 = np.array([1, 1, 1, 1], dtype=np.int32)
np.set_printoptions(precision=8, linewidth=200, suppress=True)
cudart.cudaDeviceSynchronize()
logger = trt.Logger(trt.Logger.ERROR)
builder = trt.Builder(logger)
network = builder.create_network(1 << int(trt.NetworkDefinitionCreationFlag.EXPLICIT_BATCH))</pre>
profile = builder.create_optimization_profile() # 需要使用 profile
config = builder.create_builder_config()
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (nIn, cIn, hIn, wIn))
inputT1 = network.add_input('inputT1', trt.DataType.INT32, (4, ))
inputT2 = network.add_input('inputT2', trt.DataType.INT32, (4, ))
inputT3 = network.add_input('inputT3', trt.DataType.INT32, (4, ))
profile.set_shape_input(inputT1.name, (0, 0, 0, 0), (0, 1, 1, 1), (0, 2, 2, 2)) # 这里设置的不是 shape
input 的形状而是值
profile.set_shape_input(inputT2.name, (1, 1, 1, 1), (1, 2, 3, 4), (1, 3, 4, 5))
profile.set_shape_input(inputT3.name, (1, 1, 1, 1), (1, 1, 1, 1), (1, 1, 1, 1))
config.add_optimization_profile(profile)
sliceLayer = network.add\_slice(inputT0, (0, 0, 0, 0), (0, 0, 0, 0), (0, 0, 0, 0))
#sliceLayer.set_input(0,inputT0)
sliceLayer.set_input(1, inputT1)
sliceLayer.set_input(2, inputT2)
sliceLayer.set_input(3, inputT3)
network.mark_output(sliceLayer.get_output(0))
engineString = builder.build_serialized_network(network, config)
engine = trt.Runtime(logger).deserialize_cuda_engine(engineString)
context = engine.create_execution_context()
context.set_shape_input(1, data1) # 运行时绑定真实形状张量值
context.set_shape_input(2, data2)
context.set_shape_input(3, data3)
_, stream = cudart.cudaStreamCreate()
inputH0 = np.ascontiguousarray(data0.reshape(-1))
inputH1 = np.ascontiguousarray(np.zeros([4], dtype=np.int32).reshape(-1)) # 传形状张量数据可用垃圾值
inputH2 = np.ascontiguousarray(np.zeros([4], dtype=np.int32).reshape(-1))
inputH3 = np.ascontiguousarray(np.zeros([4], dtype=np.int32).reshape(-1))
outputH0 = np.empty(context.get_binding_shape(4), dtype=trt.nptype(engine.get_binding_dtype(4)))
_, inputD0 = cudart.cudaMallocAsync(inputH0.nbytes, stream)
_, inputD1 = cudart.cudaMallocAsync(inputH1.nbytes, stream)
_, inputD2 = cudart.cudaMallocAsync(inputH2.nbytes, stream)
  inputD3 = cudart.cudaMallocAsync(inputH3.nbytes, stream)
_, outputD0 = cudart.cudaMallocAsync(outputH0.nbytes, stream)
cudart.cudaMemcpyAsync(inputD0, inputH0.ctypes.data, inputH0.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyHostToDevice, stream)
cudart.cudaMemcpyAsync(inputD1, inputH1.ctypes.data, inputH1.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyHostToDevice, stream)
cudart.cudaMemcpyAsync(inputD2, inputH2.ctypes.data, inputH2.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyHostToDevice, stream)
cudart.cudaMemcpyAsync(inputD3, inputH3.ctypes.data, inputH3.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyHostToDevice, stream)
context.execute_async_v2([int(inputD0), int(inputD1), int(inputD2), int(inputD3), int(outputD0)],
stream)
cudart.cudaMemcpyAsync(outputH0.ctypes.data, outputD0, outputH0.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyDeviceToHost, stream)
cudart.cudaStreamSynchronize(stream)
```

```
print("inputH0 :", data0.shape)
print(data0)
print("inputH1 :", data1.shape)
print(data1)
print("inputH2 :", data2.shape)
print(data2)
print("inputH3 :", data2.shape)
print(data3)
print("outputH0:", outputH0.shape)
print(outputH0)
cudart.cudaStreamDestroy(stream)
cudart.cudaFree(inputD0)
cudart.cudaFree(outputD0)
```

- 输出张量形状 (1,2,3,4), 结果与静态 set_input 示例代码相同
- 建立网络时需要 profile,并在运行时绑定真实形状张量的值,否则会有下面几种报错:

```
# 没有用 profile
[TRT] [E] 4: [network.cpp::validate::2919] Error Code 4: Internal Error (Network has dynamic or shape
inputs, but no optimization profile has been defined.)
# 没有正确设置 profile
[TRT] [E] 4: [network.cpp::validate::2984] Error Code 4: Internal Error (inputT1: optimization profile
is missing values for shape input)
# 没有在运行时绑定形状张量的值
[TRT] [E] 3: [executionContext.cpp::resolveSlots::1481] Error Code 3: API Usage Error (Parameter check
failed at: runtime/api/executionContext.cpp::resolveSlots::1481, condition:
allInputShapesSpecified(routine)
)
# 绑定的形状张量的值与 profile 不匹配
[TRT] [E] 3: [executionContext.cpp::setInputShapeBinding::1016] Error Code 3: API Usage Error (Parameter
check failed at: runtime/api/executionContext.cpp::setInputShapeBinding::1016, condition: data[i] <=</pre>
profileMaxShape[i]. Supplied binding shapes [500,500,500,500] for bindings[3] exceed min ~ max range at
index 0, maximum shape in profile is 1, minimum shape in profile is 1, but supplied shape is 500.
)
```

dynamic shape 模式下的 shuffle + set_input

```
import numpy as np
from cuda import cudart
import tensorrt as trt
nIn, cIn, hIn, wIn = 1, 3, 4, 5
10 + np.arange(wIn).reshape(1, 1, wIn)
data = data.reshape(nIn, cIn, hIn, wIn).astype(np.float32)
np.set_printoptions(precision=8, linewidth=200, suppress=True)
cudart.cudaDeviceSynchronize()
logger = trt.Logger(trt.Logger.ERROR)
builder = trt.Builder(logger)
network = builder.create_network(1 << int(trt.NetworkDefinitionCreationFlag.EXPLICIT_BATCH))</pre>
profile = builder.create_optimization_profile() # 需要使用 profile
config = builder.create_builder_config()
config.max_workspace_size = 1 << 30</pre>
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (-1, -1, -1))
```

```
profile.set_shape(inputT0.name, (1, 1, 1, 1), (nIn, cIn, hIn, wIn), (nIn * 2, cIn * 2, hIn * 2, wIn *
config.add_optimization_profile(profile)
oneLayer = network.add_constant([4], np.array([0, 1, 1, 1], dtype=np.int32))
shape0Layer = network.add_shape(inputT0)
shape1Layer = network.add_elementwise(shape0Layer.get_output(0), oneLayer.get_output(0),
trt.ElementWiseOperation.SUB)
sliceLayer = network.add_slice(inputT0, (0, 0, 0, 0), (0, 0, 0, 0), (1, 1, 1, 1)) # 给 inputT0 除了最高维
以外每一维减少 1
sliceLayer.set_input(2, shape1Layer.get_output(0))
\#shape1 = [1] + [ x-1 for x in inputT0.shape[1:] ]
#sliceLayer = network.add_slice(inputT0,(0,0,0,0),shape1,(1,1,1,1)) # 错误的做法,因为 dynamic shape 模式下
inputTO.shape 可能含有 -1, 不能作为新形状
network.mark_output(sliceLayer.get_output(0))
engineString = builder.build_serialized_network(network, config)
engine = trt.Runtime(logger).deserialize_cuda_engine(engineString)
context = engine.create_execution_context()
context.set_binding_shape(0, data.shape)
_, stream = cudart.cudaStreamCreate()
inputH0 = np.ascontiguousarray(data.reshape(-1))
outputH0 = np.empty(context.get_binding_shape(1), dtype=trt.nptype(engine.get_binding_dtype(1)))
_, inputD0 = cudart.cudaMallocAsync(inputH0.nbytes, stream)
_, outputD0 = cudart.cudaMallocAsync(outputH0.nbytes, stream)
cudart.cudaMemcpyAsync(inputD0, inputH0.ctypes.data, inputH0.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyHostToDevice, stream)
context.execute_async_v2([int(inputD0), int(outputD0)], stream)
cudart.cudaMemcpyAsync(outputH0.ctypes.data, outputD0, outputH0.nbytes,
cudart.cudaMemcpyKind.cudaMemcpyDeviceToHost, stream)
cudart.cudaStreamSynchronize(stream)
print("inputH0:", data.shape) # 只打印形状
print(data)
print("outputH0:", outputH0.shape)
print(outputH0)
cudart.cudaStreamDestroy(stream)
cudart.cudaFree(inputD0)
cudart.cudaFree(outputD0)
```

• 输出张量形状 (1,2,3,4), 结果与初始示例代码相同