# **Gather** 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
lenIndex = 3
data0 = np.arange(cIn).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) # 输入数据
data1 = np.array([1, 0, 2], dtype=np.int32) # 下标数据
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()
config.max_workspace_size = 1 << 30</pre>
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (nIn, cIn, hIn, wIn))
inputT1 = network.add_input('inputT1', trt.DataType.INT32, (len(data1), ))
gatherLayer = network.add_gather(inputT0, inputT1, 1)
network.mark_output(gatherLayer.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(data0.reshape(-1))
inputH1 = np.ascontiguousarray(data1.reshape(-1))
outputH0 = np.empty(context.get_binding_shape(2), dtype=trt.nptype(engine.get_binding_dtype(2)))
_, inputD0 = cudart.cudaMallocAsync(inputH0.nbytes, stream)
_, inputD1 = cudart.cudaMallocAsync(inputH1.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)
context.execute_async_v2([int(inputD0), int(inputD1), 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("inputH0 :", data1.shape)
print(data1)
print("outputH0:", outputH0.shape)
print(outputH0)

cudart.cudaStreamDestroy(stream)
cudart.cudaFree(inputD0)
cudart.cudaFree(inputD1)
cudart.cudaFree(outputD0)
```

• 输入张量形状 (1,3,4,5), 百位表示 C 维编号, 十位表示 H 维编号, 个位表示 W 维编号

```
[ 100. 101. 102. 103. 104. ]
                                               200. 201.
                                                          202.
                                                                203.
                                                                      204.
                                 113.
                                               210. 211. 212.
    13.
         14.
                110. 111.
                           112.
                                       114.
                                                                213.
                                                                      214.
                120.
                           122.
                                       124.
                                               220. 221.
                                                          222.
                                                                223.
         24.
                     121.
                                 123.
                                                                      224.
32.
    33.
         34.
              130.
                     131. 132.
                                133.
                                       134. | 230. 231.
                                                          232.
                                                                233.
                                                                      234.
```

• 输出张量形状 (1,3,4,5), 在次高维上按照下标张量重排顺序

```
100. 101. 102.
                    103. 104.
                                                 2.
                                                      3.
                                                            4. 7
                                                                   200.
                                                                          201.
                                                                                 202.
                                                                                        203.
                                                                                               204.
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      111.
            112.
                    113.
                           114.
                                    10.
                                          11.
                                                12.
                                                      13.
                                                           14.
                                                                   210.
                                                                          211.
                                                                                 212.
                                                                                        213.
                                                                                               214.
120.
      121.
             122.
                    123.
                           124.
                                    20.
                                          21.
                                                22.
                                                      23.
                                                           24.
                                                                   220.
                                                                          221.
                                                                                 222.
                                                                                        223.
                                                                                               224.
                    133.
                           134. \, \rfloor \, \lfloor 30. \, \rfloor
                                         31.
                                                           34.
                                                                 230.
                                                                          231.
                                                                                               234.
      131.
             132.
                                                32.
                                                      33.
                                                                                 232.
                                                                                        233.
```

• TensorRT>=8.2 后需要设置 config.max\_wokspace\_size, 否则报错

```
[TRT] [W] Skipping tactic 0 due to Myelin error: myelinTargetSetPropertyMemorySize called with invalid
memory size (0).
[TRT] [E] 10: [optimizer.cpp::computeCosts::2011] Error Code 10: Internal Error (Could not find any
implementation for node {ForeignNode[(Unnamed Layer* 0) [Gather]]}.)
```

#### axis

```
gatherLayer = network.add_gather(inputT0, inputT1, 1)
gatherLayer.axis = 0 # 重设操作的维度编号,默认值 1
```

• 指定 axis=0(在最高维上按照下标张量重排顺序),输出张量形状 (3,3,4,5)

```
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                                      0.
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                                              0.
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                                                       0.7
                                                            Γ0.
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                                0.
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                                              0.
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                                                                              0.
                 3.

√ 100.

                                    101.
                                           102.
                                                 103.
                                                        200.
                                                                      201.
                                                                             202.
                                                                                    203.
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 10.
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                                                                210. 211.
                                                                             212.
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      11.
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 20.
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                                                                                    223.
                                                                                           224.
                                                        134. ] [ 230.
30.
      31.
           32.
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                      34.
                            130.
                                   131.
                                          132.
                                                 133.
                                                                      231.
                                                                             232.
                                                                                    233.
                                                                                           234.
                                     ٢٥.
                                                       0.7
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                                [0, ]
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                                                            Γ0. 0.
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                                                   0.
                                                           0.
                                                                 0.
```

- 指定 axis=1(在次高维上按照下标张量重排顺序),输出张量形状 (1,3,4,5),结果与初始示例代码相同
- 指定 axis=2(在季高维上按照下标张量重排顺序),输出张量形状 (1,3,3,5)

• 指定 axis=3(在叔高维上按照下标张量重排顺序),输出张量形状 (1,3,4,3)

```
 \begin{bmatrix} \begin{bmatrix} 1. & 0. & 2. \\ 11. & 10. & 12. \\ 21. & 20. & 22. \\ 31. & 30. & 32. \end{bmatrix} \begin{bmatrix} 101. & 100. & 102. \\ 111. & 110. & 112. \\ 121. & 120. & 122. \\ 131. & 130. & 132. \end{bmatrix} \begin{bmatrix} 201. & 200. & 202. \\ 211. & 210. & 212. \\ 221. & 220. & 222. \\ 231. & 230. & 232. \end{bmatrix} \end{bmatrix}
```

## mode (要求 TensorRT>=8.2)

### DEFAULT 模式

```
import numpy as np
from cuda import cudart
import tensorrt as trt
nIn, cIn, hIn, wIn = 1, 3, 4, 5
lenIndex = 3
data0 = np.arange(cIn).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, 1, 2], [0, 2, -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>
config = builder.create_builder_config()
config.max_workspace_size = 1 << 30</pre>
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (nIn, cIn, hIn, wIn))
inputT1 = network.add_input('inputT1', trt.DataType.INT32, data1.shape)
gatherLayer = network.add_gather(inputT0, inputT1, 1)
gatherLayer.mode = trt.GatherMode.ND
#gatherLayer.num_elementwise_dims = 0
network.mark_output(gatherLayer.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(data0.reshape(-1))
inputH1 = np.ascontiguousarray(data1.reshape(-1))
outputH0 = np.empty(context.get_binding_shape(2), dtype=trt.nptype(engine.get_binding_dtype(2)))
_, inputD0 = cudart.cudaMallocAsync(inputH0.nbytes, stream)
_, inputD1 = cudart.cudaMallocAsync(inputH1.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)
context.execute_async_v2([int(inputD0), int(inputD1), 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("inputH0 :", data1.shape)
print(data1)
print("outputH0:", outputH0.shape)
print(outputH0)

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

- 输入张量 0 与初始示例代码相同,输入张量 1 形状 (3,2),值 [[1,0],[0,2],[2,1]],操作维度 axis=2
- 指定 mode=trt.GatherMode.DEFAUT,输出张量形状(1,3,3,2,5),按索引张量抽取指定维上所有元素

```
 \begin{bmatrix} \begin{bmatrix} 10. & 11. & 12. & 13. & 14. \\ 0. & 1. & 2. & 3. & 4. \end{bmatrix} \begin{bmatrix} 0. & 1. & 2. & 3. & 4. \\ 20. & 21. & 22. & 23. & 24. \end{bmatrix} \begin{bmatrix} 20. & 21. & 22. & 23. & 24. \\ 10. & 11. & 12. & 13. & 14. \end{bmatrix} \\ \begin{bmatrix} \begin{bmatrix} 110. & 111. & 112. & 113. & 114. \\ 100. & 101. & 102. & 103. & 104. \end{bmatrix} \begin{bmatrix} 100. & 101. & 102. & 103. & 104. \\ 120. & 121. & 122. & 123. & 124. \end{bmatrix} \begin{bmatrix} 120. & 121. & 122. & 123. & 124. \\ 110. & 111. & 112. & 113. & 114. \end{bmatrix} \end{bmatrix} \\ \begin{bmatrix} \begin{bmatrix} 210. & 211. & 212. & 213. & 214. \\ 200. & 201. & 202. & 203. & 204. \end{bmatrix} \begin{bmatrix} 200. & 201. & 202. & 203. & 204. \\ 220. & 221. & 222. & 223. & 224. \end{bmatrix} \begin{bmatrix} 220. & 221. & 222. & 223. & 224. \\ 210. & 211. & 212. & 213. & 214. \end{bmatrix} \end{bmatrix}
```

- 含义:参考 Onnx Gather 算子
  - 数据张量形状  $data[d_0, d_1, \ldots, d_r]$ ,索引张量形状  $index[a_0, a_1, \ldots, a_q]$ ,指定  $axis = p \ (0 \le p \le r)$  ,则
  - 输出张量形状为  $output[d_0,d_1,\ldots,d_{p-1},a_0,a_1,\ldots,a_q,d_{p+1},d_{p+2},\ldots,d_r]$  (p=0 时以  $a_0$  开头)
  - o 注意输出张量形状中没有了  $d_p$  这一维,相当于把 data 的这一维扩展成 index 的维度。对于 index 的每一个元素 i,都要抽取  $d_p$  维上的所有元素作为输出
  - o 命循环变量  $i_j$  满足  $0 \le i_j < a_p$ ,则计算过程可以写作(numpy 语法,左边的  $i_*$  和右边的 […] 位于  $d_p$  这一维):  $output[:,:,\ldots,:,i_0,i_1,\ldots,i_q,:,:,\ldots,:] = data[:,:,\ldots,:,[i_0,i_1,\ldots,i_q],:,:,\ldots,:]$
  - o 对于上面示例代码,就是: output[:,:, $i_0$ , $i_1$ ,:] = inputT0[:,:, $[i_0$ , $i_1$ ],:],其中  $0 \le i_0 < 4$ ,  $0 \le i_1 < 4$

#### ELEMENT 模式

```
import numpy as np
from cuda import cudart
import tensorrt as trt
nIn, cIn, hIn, wIn = 1, 3, 4, 5
lenIndex = 3
data0 = np.arange(cIn).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.zeros(data0.shape, dtype=np.int32)
np.random.seed(97)
axis = 2
# 使用随机排列
for i in range(data0.shape[0]):
    for j in range(data0.shape[1]):
        for k in range(data0.shape[3]):
            data1[i, j, :, k] = np.random.permutation(range(data0.shape[2]))
'''# 使用随机数也可以
```

```
for i in range(data0.shape[0]):
    for j in range(data0.shape[1]):
        for k in range(data0.shape[3]):
            data1[i,j,:,k] = [np.random.randint(0,data0.shape[2]) for i in range(data0.shape[2]) ]
111
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()
config.max_workspace_size = 1 << 30</pre>
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (nIn, cIn, hIn, wIn))
inputT1 = network.add_input('inputT1', trt.DataType.INT32, (nIn, cIn, hIn, wIn))
gatherLayer = network.add_gather(inputT0, inputT1, 1)
gatherLayer.mode = trt.GatherMode.ELEMENT
gatherLayer.axis = 2
network.mark\_output(gatherLayer.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(data0.reshape(-1))
inputH1 = np.ascontiguousarray(data1.reshape(-1))
outputH0 = np.empty(context.get_binding_shape(2), dtype=trt.nptype(engine.get_binding_dtype(2)))
_, inputD0 = cudart.cudaMallocAsync(inputH0.nbytes, stream)
_, inputD1 = cudart.cudaMallocAsync(inputH1.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,
\verb"cudart.cudaMemcpyKind.cudaMemcpyHostToDevice", stream")
context.execute_async_v2([int(inputD0), int(inputD1), 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("inputH0 :", data1.shape)
print(data1)
print("outputH0:", outputH0.shape)
print(outputH0)
cudart.cudaStreamDestroy(stream)
cudart.cudaFree(inputD0)
cudart.cudaFree(outputD0)
```

- 输入张量 0 与初始示例代码相同,操作维度 axis=2,输入张量 1 形状与输入张量 0 相同,值为在第 2 维上作随机排列(随机数也可以)
- 输入张量 1 形状 (1,3,4,5)

```
\left[ \left[ \begin{bmatrix} 0 & 2 & 3 & 0 & 2 \\ 3 & 3 & 2 & 3 & 3 \\ 1 & 0 & 0 & 1 & 1 \\ 2 & 1 & 1 & 2 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 3 & 3 & 2 \\ 2 & 1 & 2 & 1 & 1 \\ 3 & 2 & 0 & 2 & 0 \\ 0 & 3 & 1 & 0 & 3 \end{bmatrix} \begin{bmatrix} 0 & 2 & 2 & 2 & 1 \\ 3 & 1 & 0 & 3 & 3 \\ 2 & 3 & 3 & 0 & 2 \\ 1 & 0 & 1 & 1 & 0 \end{bmatrix} \right]
```

• 指定 mode=trt.GatherMode.ELEMENT,输出张量形状(1,3,4,5),按索引张量抽取指定位置上单个元素,表现为十位数按照输入张量 1 的次序作排列

- 含义:参考 Onnx Gather Elements 算子
  - o 数据张量、索引张量、输出张量形状相同, $data[d_0, d_1, \ldots, d_r], index[d_0, d_1, \ldots, d_r], output[d_0, d_1, \ldots, d_r],$  指定  $axis = p \ (0 \le p \le r)$
  - o 命循环变量 i 满足  $0 \le i < d_p$ ,则计算过程可以写作(numpy 语法,左边的 i 和右边的 index[...] 位于  $d_p$  这一维):  $output[:,:,\dots,:,i,:,:,\dots,:] = data[:,:,\dots,:,index[:,:,\dots,:,i,:,:,\dots,:]$
  - o 对于上面示例代码,就是: output[;;;i,:] = inputT0[;;;index[;;;i,:],:], 其中  $0 \le i < 4$

## ND 模式与 num\_elementwise\_dims 参数

```
import numpy as np
from cuda import cudart
import tensorrt as trt
nIn, cIn, hIn, wIn = 1, 3, 4, 5
lenIndex = 3
data0 = np.arange(cIn).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, 1, 2], [0, 2, -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>
config = builder.create_builder_config()
config.max_workspace_size = 1 << 30</pre>
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (nIn, cIn, hIn, wIn))
inputT1 = network.add_input('inputT1', trt.DataType.INT32, data1.shape)
gatherLayer = network.add_gather(inputT0, inputT1, 1)
gatherLayer.mode = trt.GatherMode.ND
#gatherLayer.num_elementwise_dims = 0
network.mark_output(gatherLayer.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(data0.reshape(-1))
inputH1 = np.ascontiguousarray(data1.reshape(-1))
outputH0 = np.empty(context.get_binding_shape(2), dtype=trt.nptype(engine.get_binding_dtype(2)))
_, inputD0 = cudart.cudaMallocAsync(inputH0.nbytes, stream)
_, inputD1 = cudart.cudaMallocAsync(inputH1.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)
context.execute_async_v2([int(inputD0), int(inputD1), 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("inputH0 :", data1.shape)
print(data1)
print("outputH0:", outputH0.shape)
print(outputH0)
cudart.cudaStreamDestroy(stream)
cudart.cudaFree(inputD0)
cudart.cudaFree(outputD0)
```

• 指定 mode=trt.GatherMode.ND,不指定 num\_elementwise\_dims(取默认值 0),输入张量 0 与初始示例代码相同,输入 张量 1 形状 (2,3),输出张量形状 (2,5)。索引张量从高维开始在数据张量中查找,抽取指定位置上剩余维度的所有元素

输入张量 
$$1 = \begin{bmatrix} 0 & 1 & 2 \\ 0 & 2 & -1 \end{bmatrix}$$
 输出张量  $= \begin{bmatrix} 120. & 121. & 122. & 123. & 124. \\ 230. & 231. & 232. & 233. & 234. \end{bmatrix}$ 

• 指定 mode=trt.GatherMode.ND, 指定 num\_elementwise\_dims=1, 输入张量 0 与初始示例代码相同, 输入张量 1 形状 (1,2,3), 输出张量形状 (1,2)。两个输入张量的最高 1 维必须相同,并在查找阶段跳过,索引张量从次高维开始在数据张量中查 找

data1 = np.array([[0,1,2],[0,2,-1]],dtype=np.int32)

输入张量
$$1=\begin{bmatrix}\begin{bmatrix}0&1&2\\0&2&-1\end{bmatrix}\end{bmatrix}$$
  
输出张量 $=[12.\quad 24.]$ 

• 指定 mode=trt.GatherMode.ND, 指定 num\_elementwise\_dims=2, 输入张量 0 与初始示例代码相同, 输入张量 1 形状 (1,3,2), 输出张量形状 (1,3)。两个输入张量的最高 2 维必须相同,并在查找阶段跳过,索引张量从季高维开始在数据张量中查 找

data1 = np.array([[[2,1],[3,0],[1,2]]],dtype=np.int32)

输入张量
$$1=\left[egin{bmatrix}2&1\\3&0\\1&2\end{bmatrix}
ight]$$
输出张量 $=[21.\quad130.\quad212.]$ 

• 指定 mode=trt.GatherMode.ND, 指定 num\_elementwise\_dims=3, 输入张量 0 与初始示例代码相同, 输入张量 1 形状 (1,3,4,1), 输出张量形状 (1,3,4)。两个输入张量的最高 3 维必须相同, 并在查找阶段跳过, 索引张量从叔高维开始在数据张量中查找

```
data1 = np.array([[[[0],[0],[0],[0]],[[2],[2],[2]],[[1],[1],[1],[1]]]],dtype=np.int32)
```

输入张量 
$$1=\begin{bmatrix}\begin{bmatrix}0\\0\\0\\0\end{bmatrix}\begin{bmatrix}2\\2\\2\end{bmatrix}\begin{bmatrix}1\\1\\1\\2\end{bmatrix}\end{bmatrix}$$
  
输出张量  $=\begin{bmatrix}0.&10.&20.&30.\\102.&112.&122.&132.\\201.&211.&221.&231.\end{bmatrix}$ 

- 含义:参考 Onnx GatherND 算子
  - o TensorRT 说明 <u>link</u>
  - o 数据张量形状  $data[d_0, d_1, \ldots, d_r]$ ,索引张量形状  $index[a_0, a_1, \ldots, a_q]$ ,指定  $num\_elementwise\_dims = p$ ,且  $a_q$  为 buildtime 常量,则
  - o 要求 nElementwiseDim < min(r,q), 否则报错。即要求跳过的维数不能超过 data 和 index 维数中的较小者
  - o 要求 data.shape[:nElementwiseDim] == index.shape[:nElementwiseDim],否则报错。即要求 data 和 index 的形状的 前 nElementwiseDim 维一模一样
  - o 要求 index.shape[-1] ≤ r n\*ElementwiseDim,否则报错。即要求"index 跳过 nElementwiseDim 维后的剩余维度数" (查找深度) 不能超出"data 跳过 nEementwiseDim 维后的剩余维度数"
  - o 对于第 j 维的索引  $i_j$  ( $0 \le j \le q$ ) 要求  $-d_j \le index[:,:,\dots,i_j,:,:,\dots,:] \le d_j 1$ , 即可以使用负的索引号
  - 输出张量维度 g + r index.shape[-1] 1 nElementwiseDim
  - o 记 N 为  $a_0 * a_1 * ... * a_{nElementwiseDim-1}$ , 即 data 和 index 相同维度部分的元素数
  - o If indices\_shape[-1] == r-b, since the rank of indices is q, indices can be thought of as N (q-b-1)-dimensional tensors containing 1-D tensors of dimension r-b, where N is an integer equals to the product of 1 and all the elements in the batch dimensions of the indices\_shape. Let us think of each such r-b ranked tensor as indices\_slice. Each scalar value corresponding to data[0:b-1,indices\_slice] is filled into the corresponding location of the (q-b-1)-dimensional tensor to form the output tensor
  - If indices\_shape[-1] < r-b, since the rank of indices is q, indices can be thought of as N (q-b-1)-dimensional tensor containing 1-D tensors of dimension < r-b. Let us think of each such tensors as indices\_slice. Each tensor slice corresponding to data[0:b-1, indices\_slice , :] is filled into the corresponding location of the (q-b-1)-dimensional tensor to form the output tensor
- 不满足 index.shape[-1] ≤ r n\*ElementwiseDim 时报错(报错信息的" -" 写成了 "+"):

[TRT] [E] 1: [gatherNode.cpp::computeGatherNDOutputExtents::110] Error Code 1: Internal Error (invalid dimension in GatherND indices[-1] > rank(data) + nbElementWiseDims)

• 不满足 nElementwiseDim < min(q,r) 时报错:

[TRT] [E] 3: (Unnamed Layer\* 0) [Gather]: nbElementWiseDims must between 0 and rank(data)-1 inclusive for GatherMode::kND

• 不满足 data.shape[:nElementwiseDim] == index.shape[:nElementwiseDim] 时报错:

[TRT] [E] 4: [graphShapeAnalyzer.cpp::processCheck::581] Error Code 4: Internal Error ((Unnamed Layer\*
0) [Gather]: dimensions not compatible for Gather with GatherMode = kND)