PoolingNd 层 (Pooling 层)

- 参数及示例(括号中的层名和参数名适用于 TensorRT8 及之前版本, TensorRT9 及之后被废弃)
- 初始示例代码
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- 三维池化的示例

初始示例代码

```
import numpy as np
from cuda import cudart
import tensorrt as trt
nIn, cIn, hIn, wIn = 1, 1, 6, 9 # 输入张量 NCHW
hW, wW = 2, 2 # 池化窗口 HW
data = np.tile(np.arange(1, 1 + 9, dtype=np.float32).reshape(1, 3, 3), (nIn, cIn, hIn // 3, wIn // 3))
# 输入数据
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()
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (nIn, cIn, hIn, wIn))
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (hW, wW))
network.mark_output(poolLayer.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 (1,1,6,9)

• 输出张量形状 (1,1,3,4)

$$\left[
\begin{bmatrix}
5. & 6. & 6. & 5. \\
8. & 9. & 9. & 8. \\
8. & 9. & 9. & 8.
\end{bmatrix}
\right]$$

• 计算过程:

$$\begin{bmatrix} 1 & 2 & 3 & \cdots \\ 4 & 5 & 6 & & \\ 7 & 8 & 9 & \cdots \\ \vdots & \vdots & \vdots & \end{bmatrix} = 5., \begin{bmatrix} 1 & 2 & 3 & \cdots \\ 4 & 5 & 6 & & \\ 7 & 8 & 9 & \cdots \\ \vdots & \vdots & \vdots & \end{bmatrix} = 6.$$

type

```
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (hW, wW))
poolLayer.type = trt.PoolingType.AVERAGE
```

• 指定平均值池化,输出张量形状 (1,1,3,4)

$$\left[\begin{bmatrix} 2. & 2.5 & 3. & 2. \\ 3.5 & 4. & 4.5 & 3.5 \\ 5. & 5.5 & 6. & 5. \end{bmatrix} \right]$$

• 可用的池化方式

trt.PoolingType 名	说明
AVERAGE	均值池化
MAX	最大值池化
MAX_AVERAGE_BLEND	混合池化,见下面的 blend_factor 部分

blend_factor

```
bF = 0.5
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX_AVERAGE_BLEND, (hW, wW))
poolLayer.blend_factor = bF # 均值池化的比例,默认值 1.0
```

• 指定 blend_factor=0.5,输出张量形状 (1,1,3,4)

$$\left[\left[\begin{bmatrix} 4. & 4.75 & 5. & 4. \\ 6.25 & 7. & 7.25 & 6.25 \\ 7. & 7.75 & 8. & 7. \end{bmatrix} \right] \right]$$

• 计算过程:

$$bF \cdot \left[\left[\begin{bmatrix} 3. & 3.5 & 4. & 3. \\ 4.5 & 5. & 5.5 & 4.5 \\ 6. & 6.5 & 7. & 6. \end{bmatrix} \right] \right] + (1 - bF) \cdot \left[\left[\begin{bmatrix} 5. & 6. & 6. & 5. \\ 8. & 9. & 9. & 8. \\ 8. & 9. & 9. & 8. \end{bmatrix} \right] \right] = \left[\left[\begin{bmatrix} 4. & 4.75 & 5. & 4. \\ 6.25 & 7. & 7.25 & 6.25 \\ 7. & 7.75 & 8. & 7. \end{bmatrix} \right] \right]$$

window_size_nd (window_size)

```
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (1, 1)) # 默认池化窗口大小和步长都设为给定参数 poolLayer.window_size_nd = (hW, WW) # 池化窗口大小
```

• 输出张量形状 (1,1,5,8), 池化窗口尺寸不变, 但是步长会保持 (1,1)

$$\left[\begin{bmatrix} 4. & 5. & 5. & 4. & 5. & 5. & 4. & 5. \\ 7. & 8. & 8. & 7. & 8. & 8. & 7. & 8. \\ 7. & 8. & 8. & 7. & 8. & 8. & 7. & 8. \\ 4. & 5. & 5. & 4. & 5. & 5. & 4. & 5. \\ 7. & 8. & 8. & 7. & 8. & 8. & 7. & 8. \end{bmatrix} \right]$$

• 使用旧版 API window_size 会收到警告

DeprecationWarning: Use window_size_nd instead.

stride_nd (stride)

```
hS = wS = 1
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (hW, wW))
poolLayer.stride_nd = (hS, wS) # 池化窗口步长
```

• 指定 stride=(hS,wS),输出张量形状 (1,1,5,8),

$$\left[\begin{bmatrix} 5. & 6. & 6. & 5. & 6. & 6. & 5. & 6. \\ 8. & 9. & 9. & 8. & 9. & 9. & 8. & 9. \\ 8. & 9. & 9. & 8. & 9. & 9. & 8. & 9. \\ 5. & 6. & 6. & 5. & 6. & 6. & 5. & 6. \\ 8. & 9. & 9. & 8. & 9. & 9. & 8. & 9. \end{bmatrix} \right]$$

• 使用旧版 API stride 会收到警告

DeprecationWarning: Use stride_nd instead.

padding_nd (padding)

```
hP = wP = 1
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (hW, wW))
poolLayer.padding_nd = (hP, wP) # 四周填充 0 层数,默认值 (0,0)
```

• 指定 padding=(1,1)(HW 维均填充 1 层 0) ,输出张量形状 (1,1,4,5)

• 指定 padding=(1,0)(H 维填充 1 层 0) ,输出张量形状 (1,1,4,4)

$$\left[
\begin{bmatrix}
2. & 3. & 3. & 2. \\
8. & 9. & 9. & 8. \\
5. & 6. & 6. & 5. \\
8. & 9. & 9. & 8.
\end{bmatrix}
\right]$$

• 指定 padding=(0,1)(W 维填充 1 层 0) ,输出张量形状 (1,1,3,5)

$$\left[
\begin{bmatrix}
4. & 6. & 5. & 6. & 6. \\
7. & 9. & 8. & 9. & 9. \\
7. & 9. & 8. & 9. & 9.
\end{bmatrix}
\right]$$

• 使用旧版 API padding 会收到警告

```
DeprecationWarning: Use padding_nd instead.
```

pre_padding

```
hPre = wPre = 1
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (hW, wW))
poolLayer.pre_padding = (hPre, wPre) # 头部填充 0 层数,默认值 (0,0)
```

• 指定 pre_padding=(1,1)(HW 维头部均填充 1 层 0),输出张量形状 (1,1,3,5)

$$\begin{bmatrix}
1. & 3. & 2. & 3. & 3. \\
7. & 9. & 8. & 9. & 9. \\
4. & 6. & 5. & 6. & 6.
\end{bmatrix}$$

• 指定 pre_padding=(1,0)(H 维头部填充 1 层 0),输出张量形状(1,1,3,4)

$$\left[\left[\begin{bmatrix} 2. & 3. & 3. & 2. \\ 8. & 9. & 9. & 8. \\ 5. & 6. & 6. & 5. \end{bmatrix} \right] \right]$$

• 指定 pre_padding=(0,1)(W 维头部填充 1 层 0),输出张量形状(1,1,3,5)

$$\left[
\begin{bmatrix}
4. & 6. & 5. & 6. & 6. \\
7. & 9. & 8. & 9. & 9. \\
7. & 9. & 8. & 9. & 9.
\end{bmatrix}
\right]$$

post_padding

```
hPost = wPost = 1
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (hW, wW))
poolLayer.post_padding = (hPost, wPost) # 尾部填充 0 层数, 默认值 (0,0)
```

• 指定 post_padding=(1,1)(HW 维尾部均填充 1 层 0),输出张量形状(1,1,3,5)

$$\left[\left[\begin{bmatrix} 5. & 6. & 6. & 5. & 6. \\ 8. & 9. & 9. & 8. & 9. \\ 8. & 9. & 9. & 8. & 9. \end{bmatrix} \right]$$

• 指定 post padding=(1,0)(H 维尾部填充 1 层 0),输出张量形状 (1,1,3,4)

• 指定 post_padding=(0,1)(W 维尾部填充 1 层 0),输出张量形状 (1,1,3,5)

padding_mode

```
inputT0 = network.add_input('inputT0', trt.DataType.FLOAT, (nIn, cIn, hIn - 1, wIn)) # 去除输入张量的最后一行,以便观察结果
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (hW, wW))
poolLayer.padding_mode = trt.PaddingMode.SAME_UPPER
```

- 计算过程参考 TensorRT C API reference
- 指定 padding_mode = trt.PaddingMode.SAME_UPPER,输出张量形状 (1,1,3,5)

• 指定 padding_mode = trt.PaddingMode.SAME_LOWER, 输出张量形状 (1,1,3,5)

$$\left[\begin{bmatrix} 1. & 3. & 2. & 3. & 3. \\ 7. & 9. & 8. & 9. & 9. \\ 4. & 6. & 5. & 6. & 6. \end{bmatrix} \right]$$

• 指定 padding_mode = trt.PaddingMode.EXPLICIT_ROUND_UP,输出张量形状 (1,1,3,5)

• 指定 padding_mode = trt.PaddingMode.EXPLICIT_ROUND_DOWN,输出张量形状 (1,1,2,4)

$$\left[\left[\begin{bmatrix} 5. & 6. & 6. & 5. \\ 8. & 9. & 9. & 8. \end{bmatrix} \right] \right]$$

• 指定 padding mode = trt.PaddingMode.CAFFE ROUND UP, 输出张量形状 (1,1,3,5)

$$\left[\begin{bmatrix} 5. & 6. & 6. & 5. & 6. \\ 8. & 9. & 9. & 8. & 9. \\ 5. & 6. & 6. & 5. & 6. \end{bmatrix} \right]$$

• 指定 padding_mode = trt.PaddingMode.CAFFE_ROUND_DOWN,输出张量形状 (1,1,2,4)

```
\left[ \left[ \begin{bmatrix} 5. & 6. & 6. & 5. \\ 8. & 9. & 9. & 8. \end{bmatrix} \right] \right]
```

average_count_excludes_padding

```
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.AVERAGE, (hW, wW))
poolLayer.padding_nd = (1, 1) # 不支持 pre_padding 或 post_padding
poolLayer.average_count_excludes_padding = False # 是否排除分母中填充 0 的计数,默认值 True
```

• 指定 average_count_excludes_padding=False,均值池化连着填充的 0 一起计算(默认填充的 0 不计入分母),输出张量形状 (1,1,4,5)

```
\left[ \begin{bmatrix} 0.25 & 1.25 & 0.75 & 1. & 1.25 \\ 2.75 & 7. & 6. & 6.5 & 7. \\ 1.25 & 4. & 3. & 3.5 & 4. \\ 1.75 & 4.25 & 3.75 & 4. & 4.25 \end{bmatrix} \right]
```

三维池化的示例

```
import numpy as np
from cuda import cudart
import tensorrt as trt
nIn, cIn, hIn, wIn = 1, 2, 6, 9 # 输入张量 NCHW
cW, hW, wW = 2, 2, 2 # 池化窗口 HW
data = np.tile(np.arange(1, 1 + 9, dtype=np.float32).reshape(3, 3), (2, 2, 3)).reshape(nIn, 1, cIn, hIn,
wIn)
data[0, 0, 1] *= 10
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, 1, cIn, hIn, wIn))
poolLayer = network.add_pooling_nd(inputT0, trt.PoolingType.MAX, (cW, hW, wW))
network.mark_output(poolLayer.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 (1,1,2,6,9)

```
Г10. 20. 30. 10.
                                                            20. 30. 10. 20.
                                                                                 30.
     2. 3. 1. 2. 3. 1. 2. 3.
                                        40. 50.
                                                  60.
                                                       40.
                                                            50.
                                                                      40.
                                                                                 60.
                                                                 60.
                                                                            50.
7. 8. 9. 7. 8. 9. 7. 8. 9. 1. 2. 3. 1. 2. 3. 4. 5. 6. 4. 5. 6. 4. 5. 6.
                                       70. 80.
                                                  90.
                                                       70.
                                                            80.
                                                                 90.
                                                                      70. 80.
                                                                                 90.
                                       10. 20.
                                                  30. 10.
                                                            20.
                                                                                 30.
                                                                 30.
                                                                      10.
                                                                            20.
                                       40. 50.
                                                  60.
                                                                                 60.
                                                       40.
                                                            50.
                                                                 60.
                                                                      40.
                                                                            50.
                                     L 70.
                                            80.
                                                  90. 70.
                                                            80.
                                                                 90.
                                                                      70.
                                                                            80.
                                                                                 90.
```

• 输出张量形状 (1,1,1,3,4), 最大元素全都来自靠后的通道

```
\left[ \left[ \left[ \begin{array}{cccc} 50. & 60. & 60. & 50. \\ 80. & 90. & 90. & 80. \\ 80. & 90. & 90. & 80. \end{array} \right] \right] \right]
```