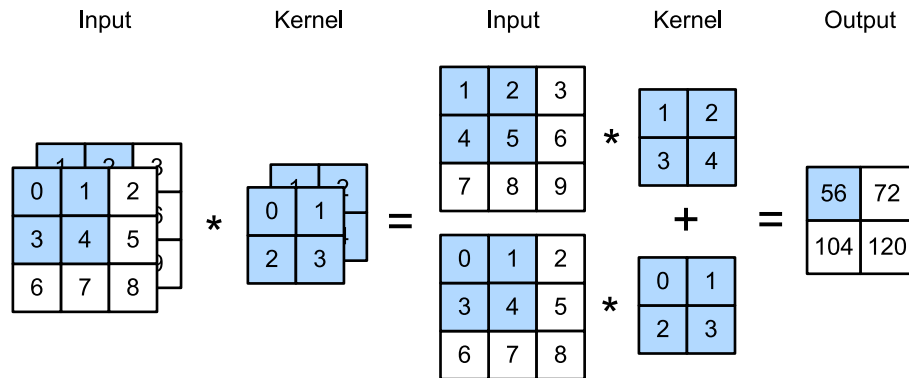


Multiple Input and Output Channels

Multiple Input Channels



```
In [1]: import d2l
        from mxnet import nd

        def corr2d_multi_in(X, K):
            # First, traverse along the 0th dimension (channel dimension) of X and K.
            # Then, add them together by using *
            return nd.add_n(*[d2l.corr2d(x, k) for x, k in zip(X, K)])
```

We can construct the input array X and the kernel array K of the above diagram to validate the output of the cross-correlation operation.

```
In [2]: X = nd.array([[[0, 1, 2], [3, 4, 5], [6, 7, 8]],  
                      [[1, 2, 3], [4, 5, 6], [7, 8, 9]]])  
K = nd.array([[[0, 1], [2, 3]], [[1, 2], [3, 4]]])  
  
corr2d_multi_in(X, K)
```

```
Out[2]: [[ 56.  72.]  
         [104. 120.]]  
<NDArray 2x2 @cpu(0)>
```

Multiple Output Channels

For multiple output channels we simply generate multiple outputs and then stack them together.

```
In [3]: def corr2d_multi_in_out(X, K):  
        # Traverse along the 0th dimension of K, and each time, perform cross-correlation  
        # operations with input X. All of the results are merged together using the stack function.  
        return nd.stack(*[corr2d_multi_in(X, k) for k in K])
```

We construct a convolution kernel with 3 output channels by concatenating the kernel array K with $K+1$ (plus one for each element in K) and $K+2$.

```
In [4]: K = nd.stack(K, K + 1, K + 2)  
        K.shape
```

```
Out[4]: (3, 2, 2, 2)
```

We can have multiple input and output channels.

```
In [5]: print(X.shape)
        print(K.shape)
        print(corr2d_multi_in_out(X, K))
```

```
(2, 3, 3)
```

```
(3, 2, 2, 2)
```

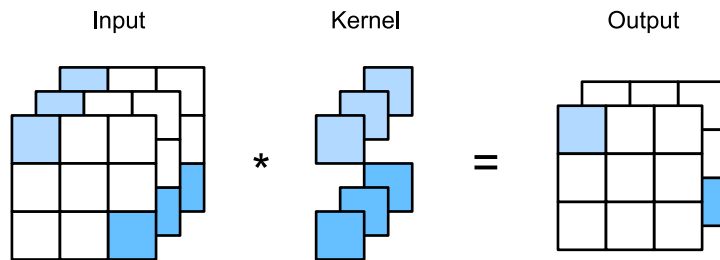
```
[[[ 56.  72.]
    [104. 120.]]
```

```
[[ 76. 100.]
 [148. 172.]]
```

```
[[ 96. 128.]
 [192. 224.]]]
```

```
<NDArray 3x2x2 @cpu(0)>
```

1 × 1 Convolutions



```
In [6]: def corr2d_multi_in_out_1x1(X, K):  
        c_i, h, w = X.shape  
        c_o = K.shape[0]  
        X = X.reshape((c_i, h * w))  
        K = K.reshape((c_o, c_i))  
        Y = nd.dot(K, X) # Matrix multiplication in the fully connected layer.  
        return Y.reshape((c_o, h, w))
```

This is equivalent to cross-correlation with an appropriately narrow 1×1 kernel.

```
In [7]: X = nd.random.uniform(shape=(3, 3, 3))
        K = nd.random.uniform(shape=(2, 3, 1, 1))

        Y1 = corr2d_multi_in_out_1x1(X, K)
        Y2 = corr2d_multi_in_out(X, K)

        (Y1 - Y2).norm().asscalar() < 1e-6
```

```
Out[7]: True
```