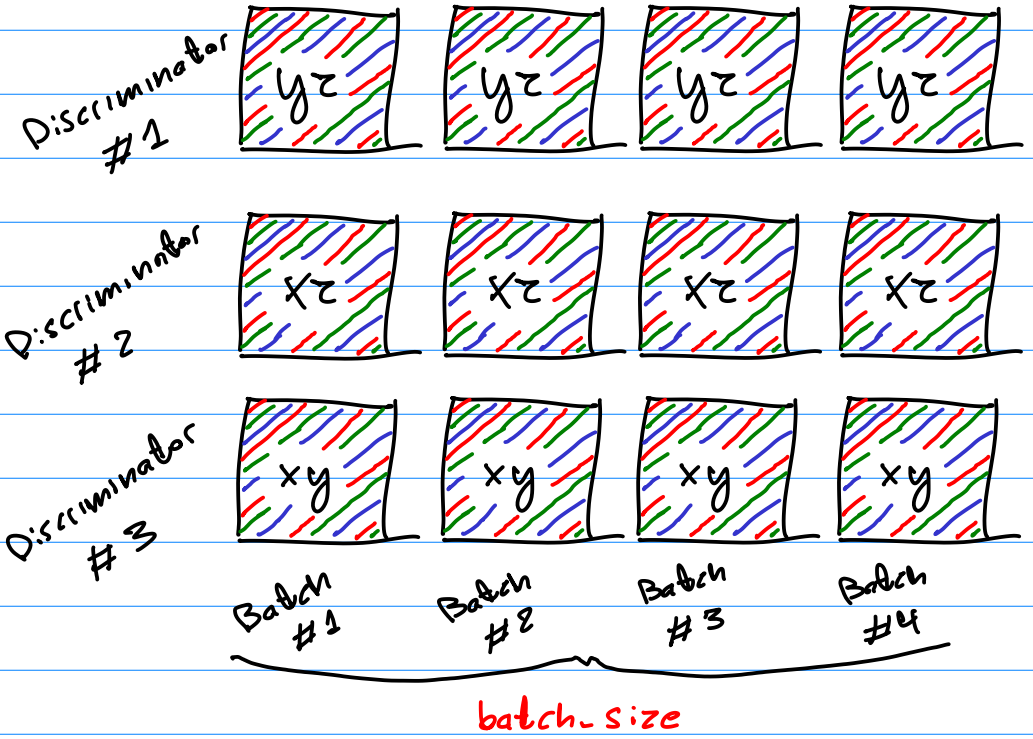
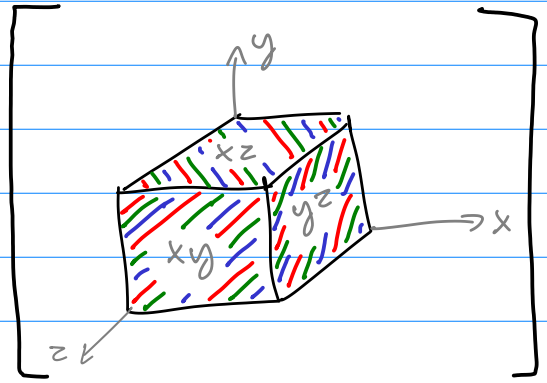
Real



Fake



$\times$ [D_batch_size]

The Fake data exists as a 5^{th} dimensional tensor:

[D_batch_size, nc, z, y, x]

- x, y, z → dims (image size, same in all directions)
- nc → number of channels (Quats 1-4, FeatureIds, etc...)
- D_batch_size → Number of generated 3-D microstructures

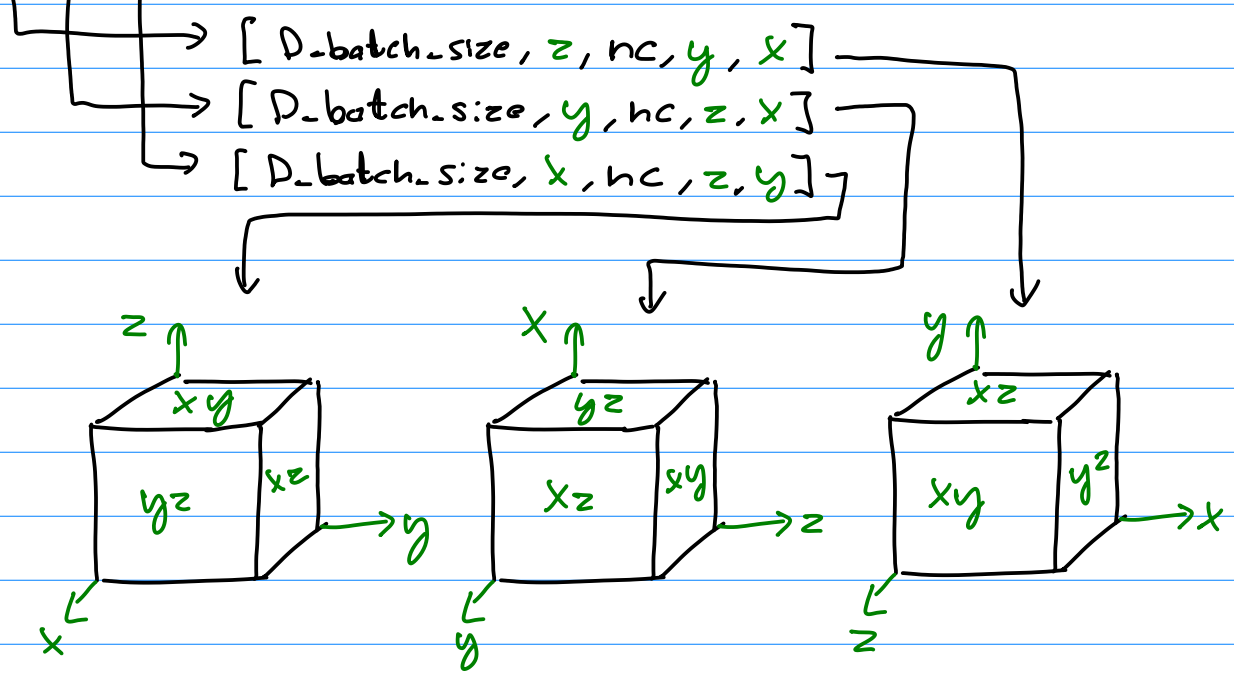
To use this 5-D tensor in a discriminator that can only parse 4-D tensors, it must be reshaped...

```
data_fake_perm = data_fake.permute(*c_perm_dim).reshape(*shape_disc)
```

① ②

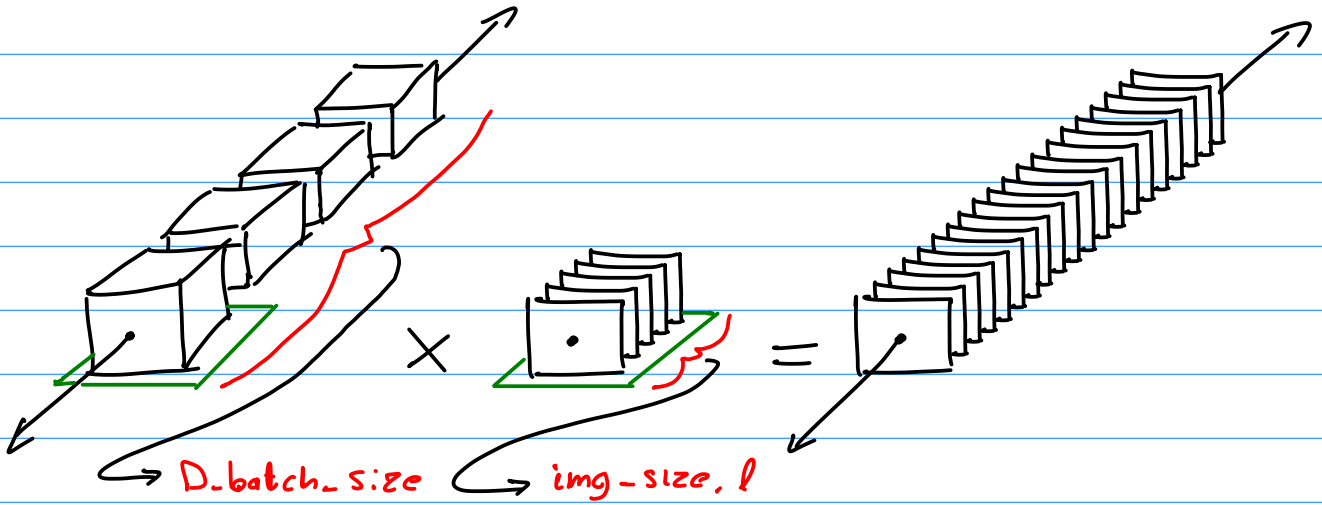
① Reorganize and Rotate the data cube

```
c_perm = [
    [0, 2, 1, 3, 4], # last 3 indecies are now [nc,y,x] -> z normal
    [0, 3, 1, 2, 4], # last 3 indecies are now [nc,z,x] -> y normal
    [0, 4, 1, 2, 3]  # last 3 indecies are now [nc,z,y] -> x normal
```



② Reshape the tensor

- shape_disc = [l * D_batch_size, nc, l, l]
  # combine batch size and slice normal plane
  # (turn the entire unused direction into batches)
- X, y, and z dimensions all have the same length (image size [l])
- By reshaping l x D_batch_size, sliceGan is lining up the data cubes along the face of interest's normal vector:



Each rotate + permute operation yields a large batch of 2-D data as a 4-D tensor

img_size (l) = 64

# Inputs		# outputs		
batch_size	D_batch_size	real	fake	
8	2	8	128	✓
16	4	16	256	✓
64	1	64	64	✓