

Difference between OS and Inception

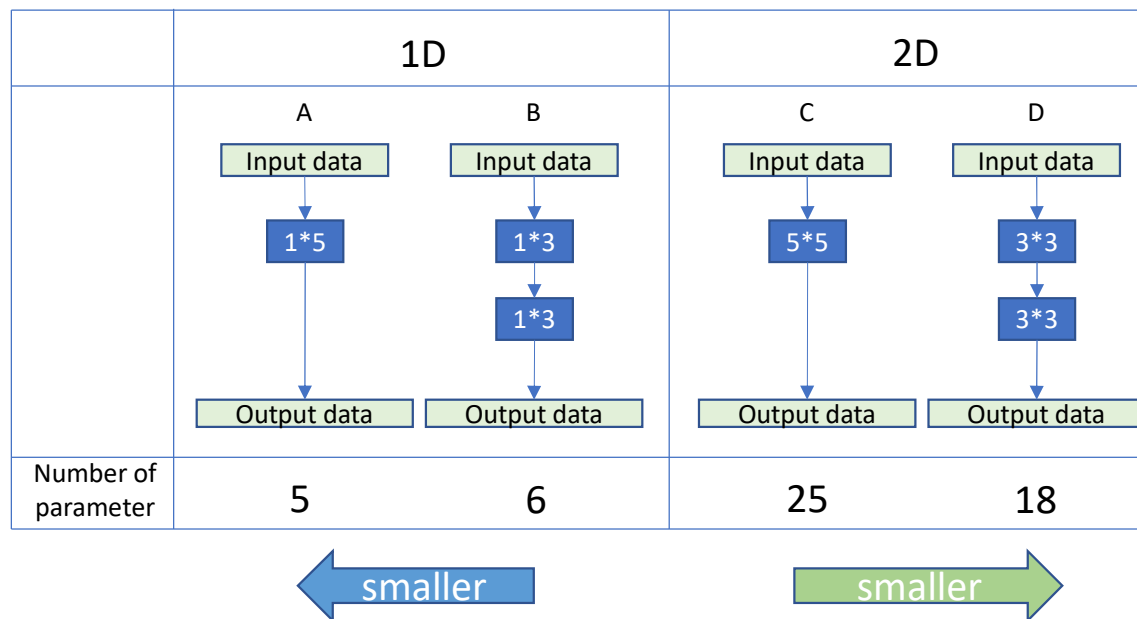
You might think OS design is a type of Inception.
It is true. Omin-scale is a type of multi-scale.
However, there are something behind it.

To summarise, OS is a **shallower** design (it should be)
OS is built by a **unique** model size reduction method.

1D-CNN tends to be shallower

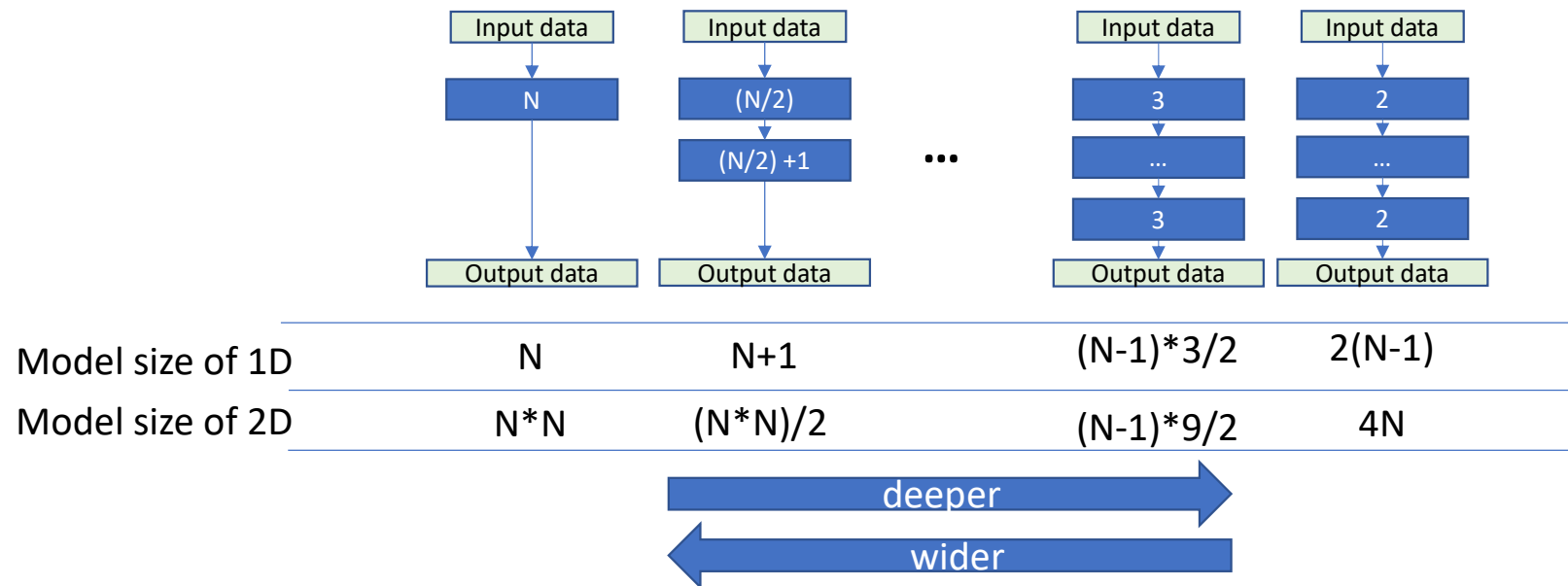
Start from a simple case

- In 1D-CNN **kernel factorization** cannot reduce mode size as 2D-CNN. It will increase the model size.



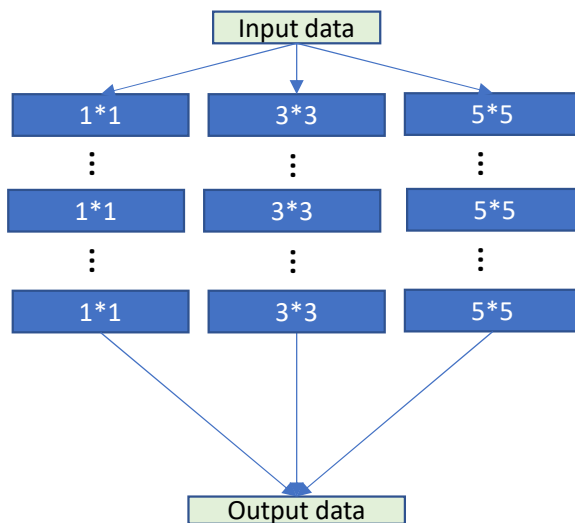
Supposing the ideal size is N

- For 1D-CNN, going deeper will **increase** the model size
- For 2D-CNN, going deeper will **reduce** the model size



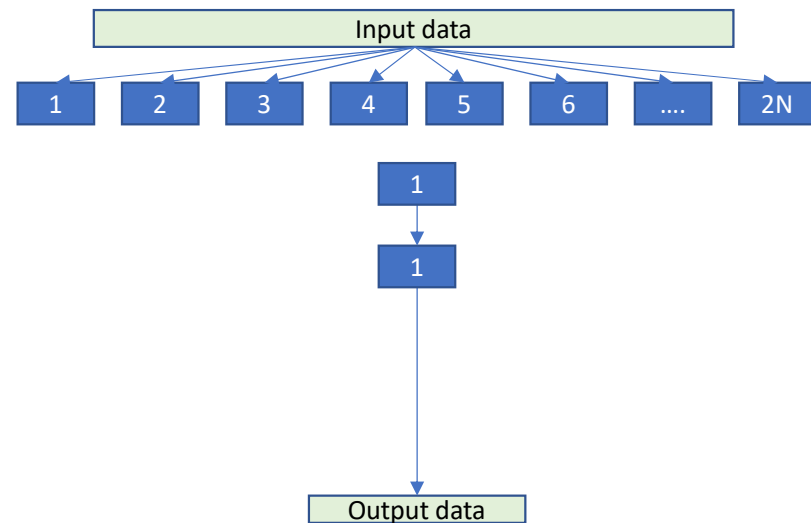
For smaller model size

2D-CNN should be
Stack of **many layers** of **small kernels** to
build RF of different sizes



Inception

1D-CNN should be
Stack of **few layers** of **various kernels** to
have RF of different size
Following layer is used for add non-linear



All-RF design

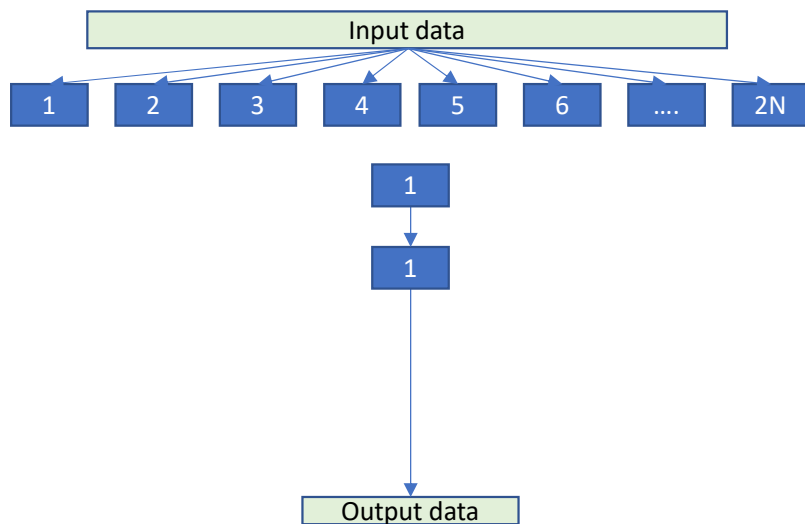
The First difference of inception and OS

- For 2D-CNN, the smaller model size solution to build RF of different size is:
 - **stacking** many thinner layer to build RF of difference sizes.
 - (stacking will make it **deeper**)
- For 1D-CNN, the smaller model size solution to build RF of different size is:
 - Just using RF of difference sizes.
 - (just using will make it **shallower**)

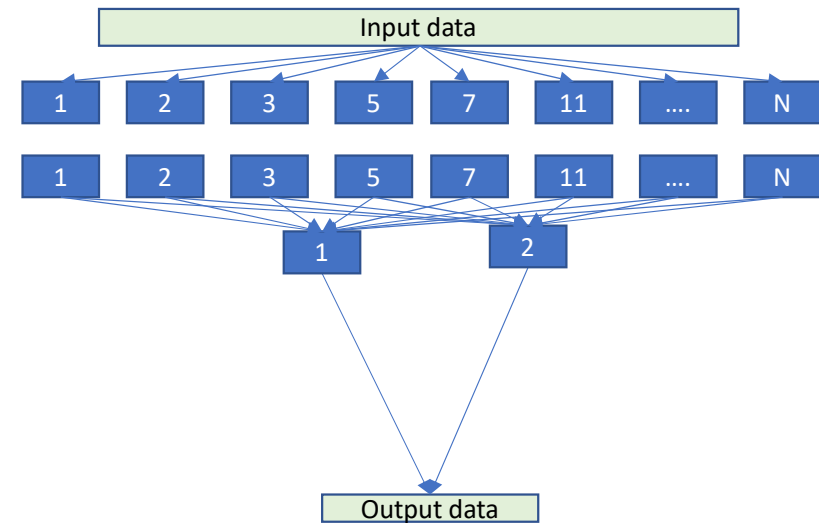
OS is built by a **unique** model size
reduction method.

Can we make the model size more smaller?

All-RF



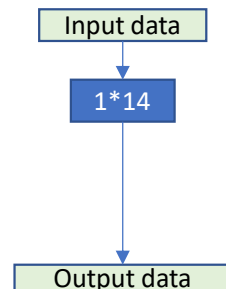
OS-CNN



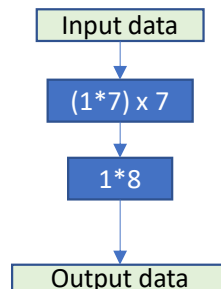
In order to learn same information as the All-RF one
The OS-CNN need to have **more channel number** for each
kernel. Therefore, we cannot merely calculate mode size by
sum the kernel size.

What's the channel number should be?

- As the subsection “No representation ability lose”[1][2] says
 - If we want to do kernel factorization we need to increase the channel number of first layer
 - And the channel number is $\min(A, B)$. Where a and b are kernel sizes. The example is below:



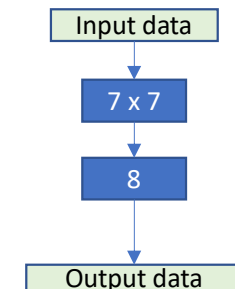
A



B

From A to B the channel requirement for $1*7$ kernel is 7

Let's write it in and remove $1*$ for the following discussion is all based on 1D-CNN. Then we should write B as C

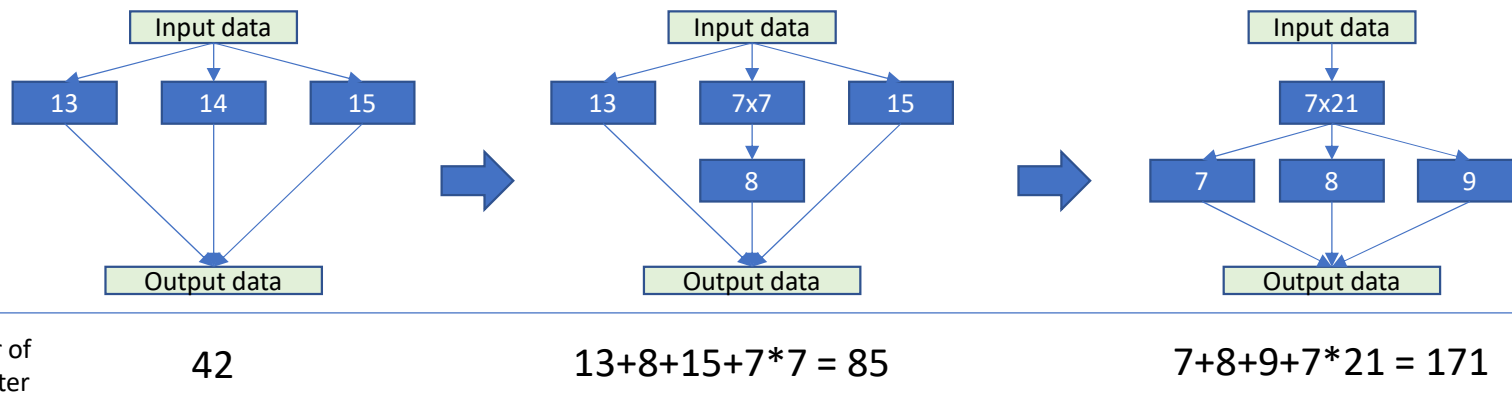


C

[1] https://github.com/Wensi-Tang/OS-CNN/blob/master/Code_example_of_theoretical_proof/4_3_Check_No_representation_ability_lose.ipynb

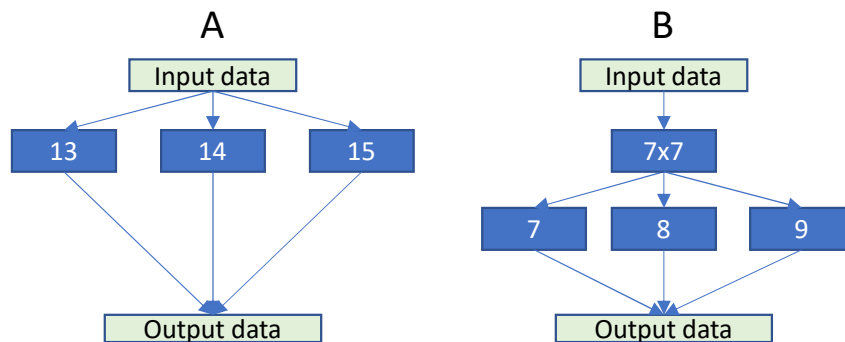
[2] <https://github.com/Wensi-Tang/OS-CNN/blob/master/Appendix/Proof%20of%20No%20representation%20ability%20lose.pdf>

Simple factorization will still increase model size



However

- The objective that we want kernels of all size is:
 - we want to find the proper kernel.
 - Therefore, we don't really need 21 channels.
- For example, if we just want 1 kernel. Then, the model should be selected from {13, 14, 15} during training time, and model B should be selected from {7, 8, 9}. And for network B, only 7 channels would be enough!



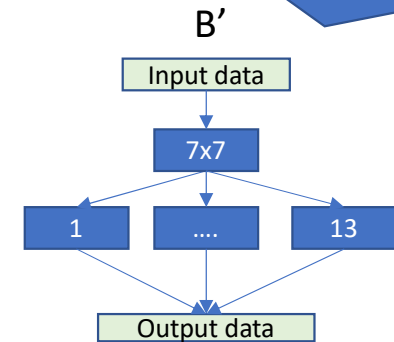
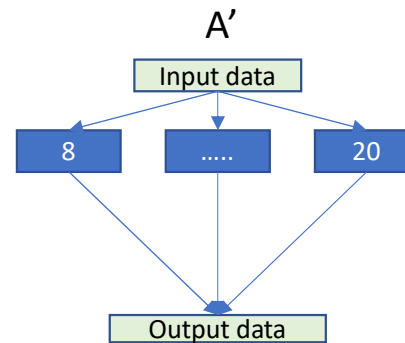
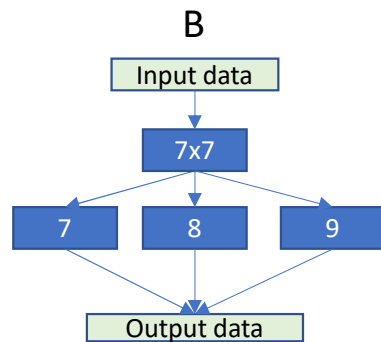
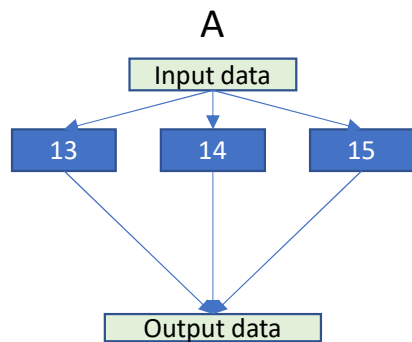
Number of
parameter

42

$7+8+9+7*7 = 73$

It seems A is still of smaller number of parameter than B, But.....

If we don't want to select from {13, 14, 15}, if we want to select from 8 to 20



This is of smaller number of parameter!

Number of parameter

42

$7+8+9+7*7 = 73$

$8+9+...+20 = 182$

$(1+2,...,+13)+49 = 140$

The Second difference of OS and inception

- This kind of model size reduction does not even exist in 2D-CNN.

It seems A is still of smaller number of parameter than B, But.....

