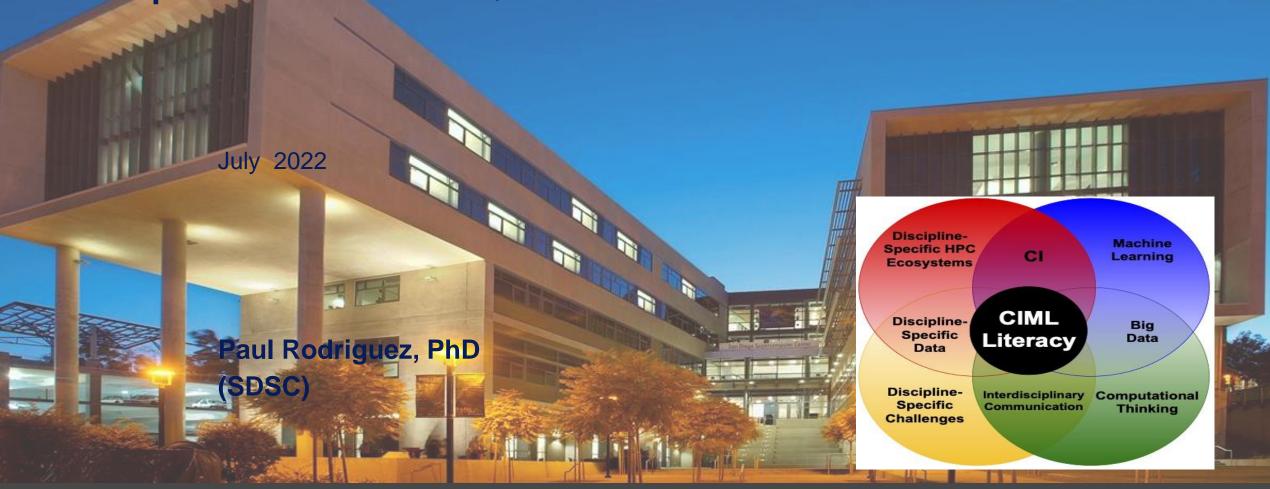
## Deep Learning Topics: Skip Connections, and the Keras Model API



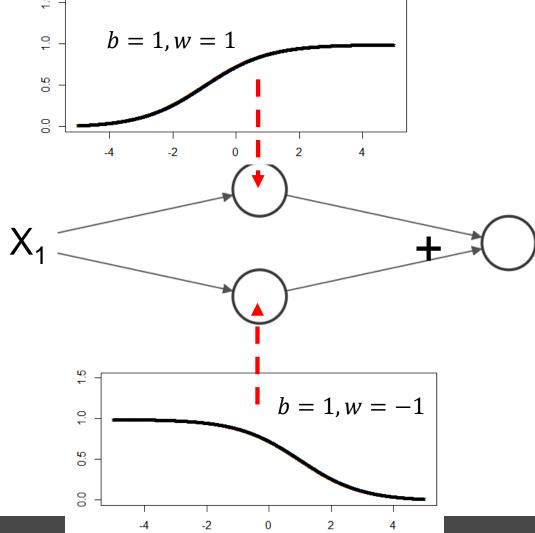


#### **Outline**

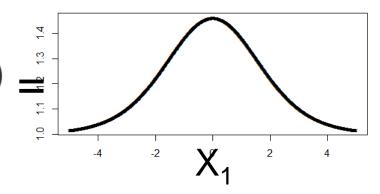
- Gate connection idea
- Skip and Residual connections
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### Recall: the logistic unit

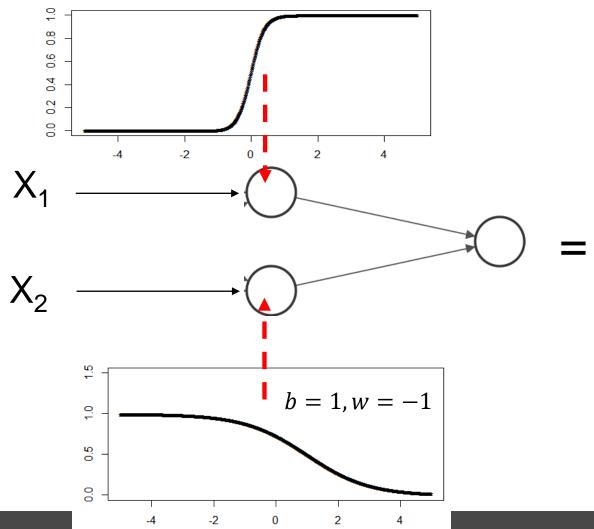
## **Example: 1 input into 2 logistic units with these activations**



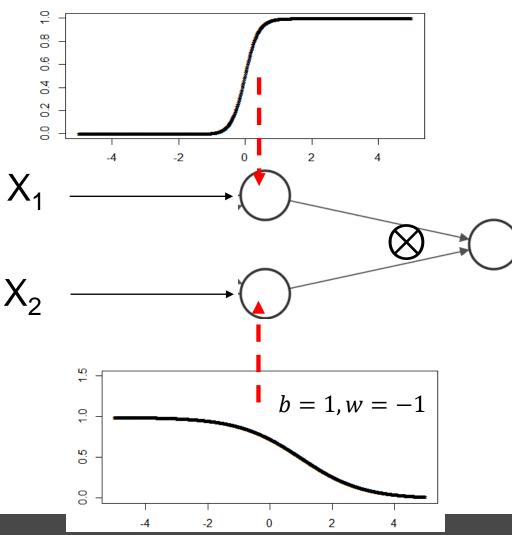
If you add these 2 units into a final output unit what would the output function look like?



## Example: 2 input into 2 logistic units with these activations

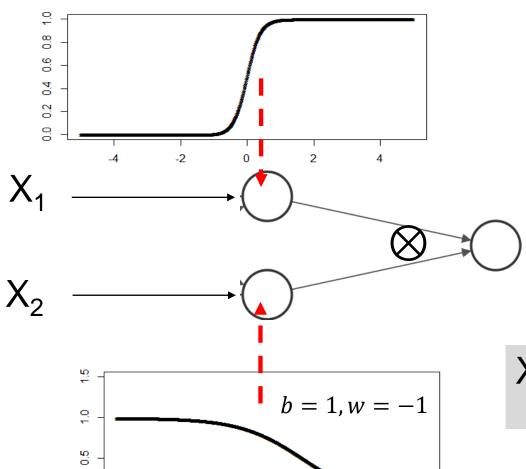


## **Example: 2 input into 2 logistic units with these activations**



What if you multiply these?
What is the output function doing?

## **Example: 2 input into 2 logistic units with these activations**



What if you multiply these?
What is the output function doing?

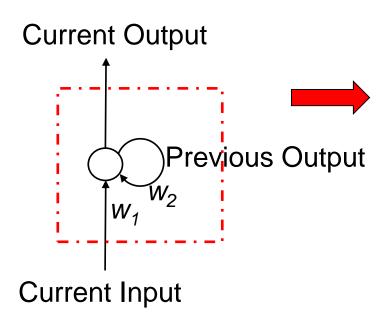
$$= \begin{cases} 0 & \text{if } X_1 < 0 \\ h(X_2) & \text{if } X_1 > 0 \end{cases}$$

X<sub>1</sub>"gates" X<sub>2</sub> activation

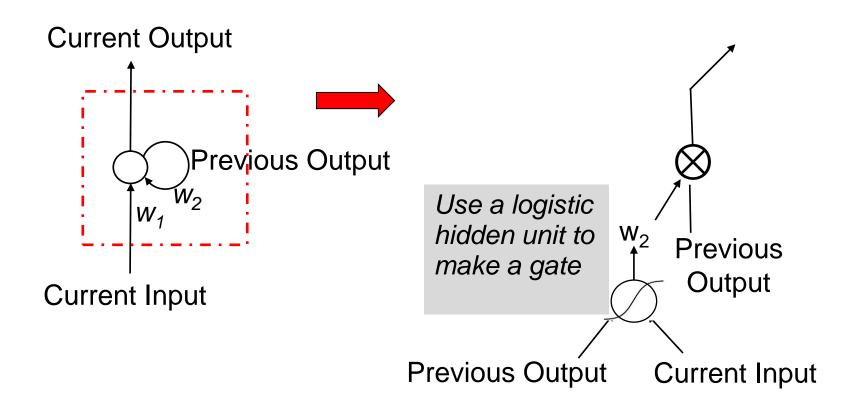
2

-2

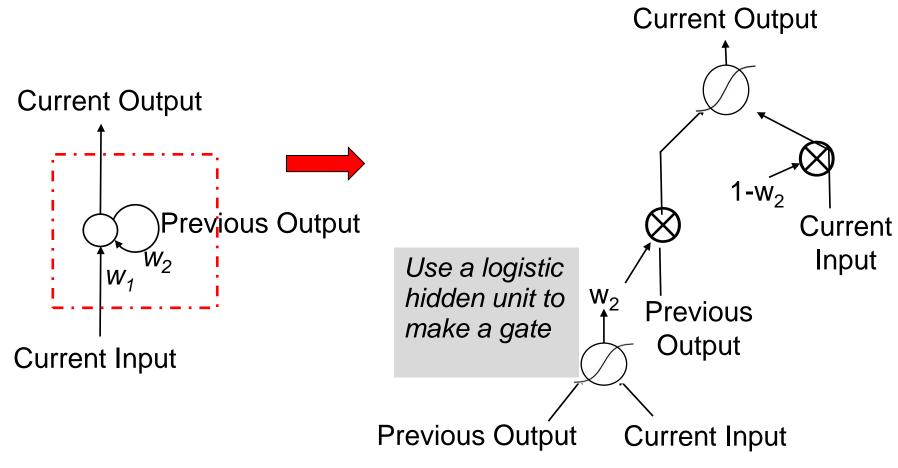
## A recurrent unit for sequence learning can be replaced by a gated unit



## A recurrent unit for sequence learning can be replaced by a gated unit



## A recurrent unit for sequence learning can be replaced by a gated unit

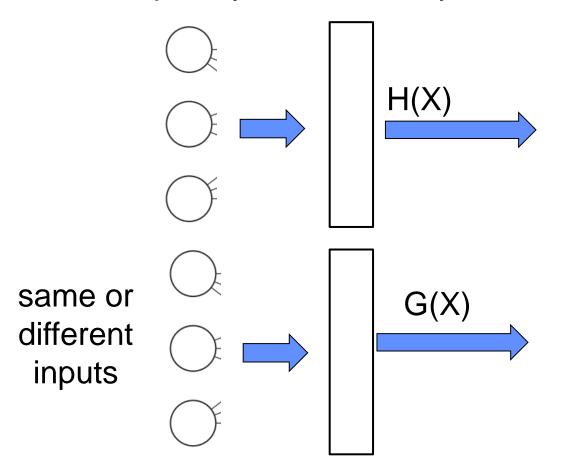


Use the gate to either keep previous output or update it with current input

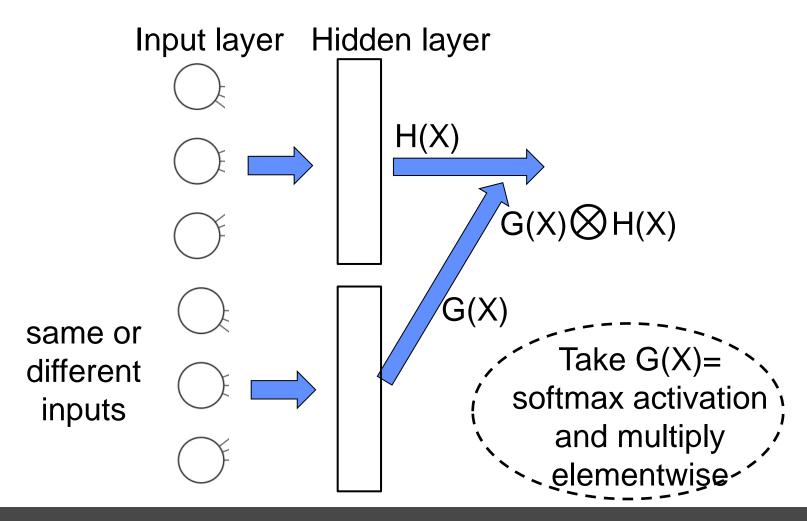
'Gated Recurrent Unit' Cho, Bengio 2015

### Redrawing the gate for two sets of hidden units

Input layer Hidden layer



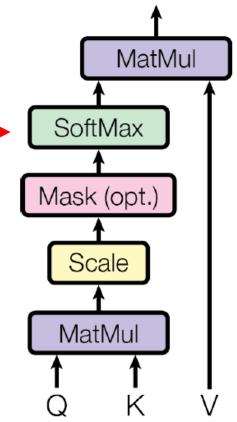
## Use softmax for G(X) to get gating weights



Recall: softmax normalizes outputs into probability weights

## Scaled Dot-Product Attention (very rough summary)

"Attention" mechanism in language transformers use a softmax gate



The gate is applied to possible Values (V) for decoding

English vs French Words 'align' to make a gate

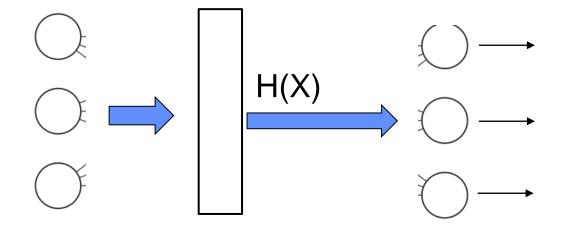
Vaswani, et al. 2017 Attention Is All You Need (for Transformers)

#### **Outline**

- Gate connection idea
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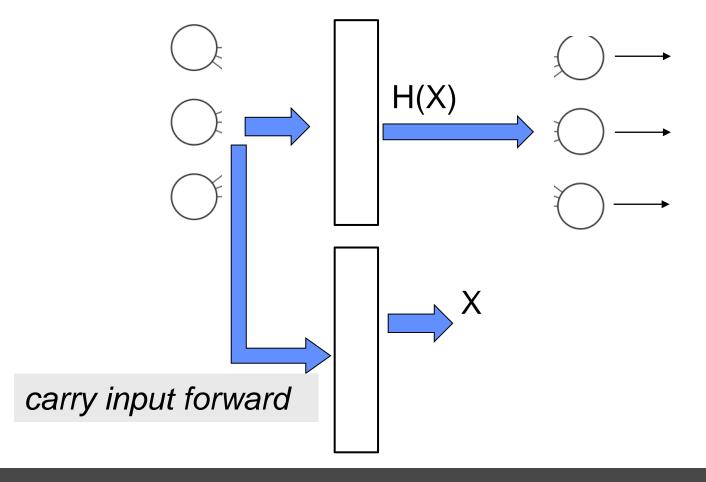
## Recall the Multilayer Perceptron (MLP)

Input layer Hidden layer Output layer

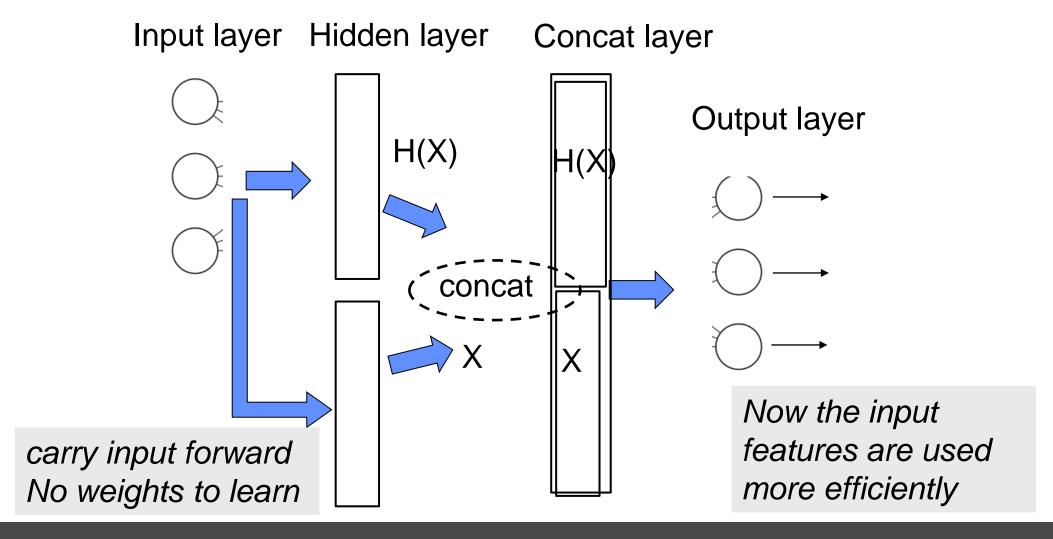


# To help the MLP learn directly from input carry input forward

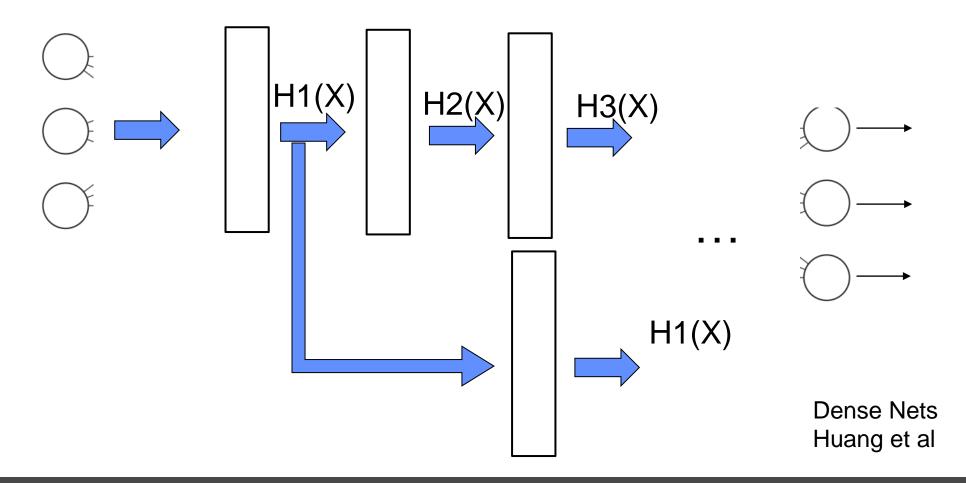
Input layer Hidden layer Output layer



## Concatenate input with hidden units into new layer

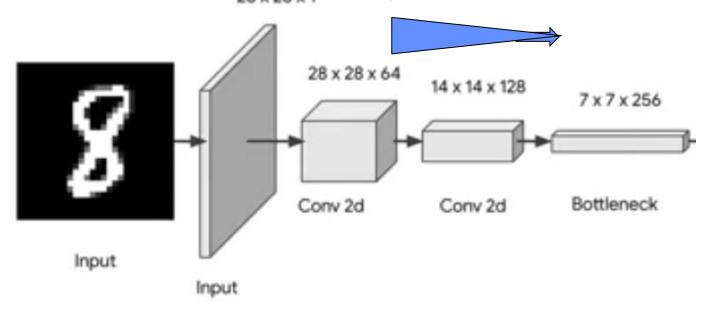


## Can be done for any (or all) previous layer and skip any number of layers



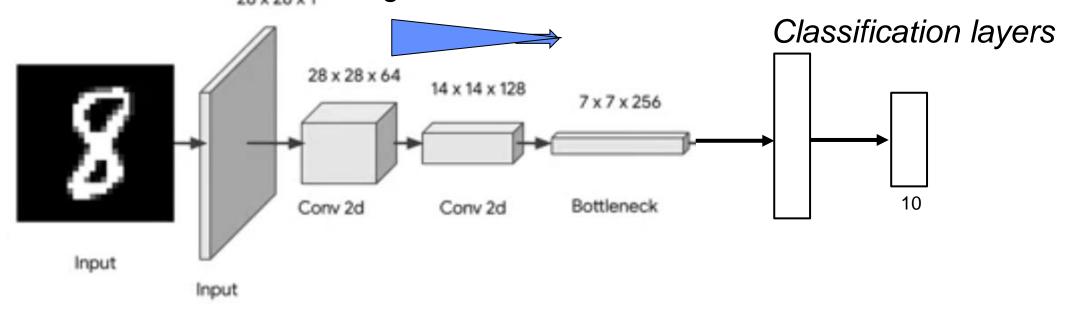
#### Recall: CNN architecture for MNIST classification

more feature maps & downsampling : 'encoding' features



#### Consider: CNN architecture for MNIST classification

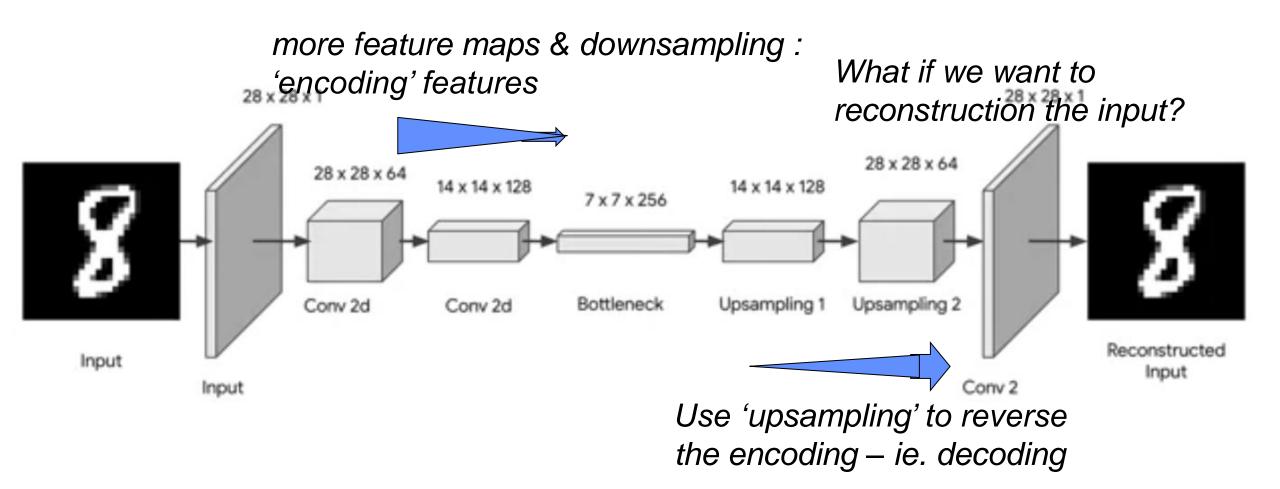
more feature maps & downsampling : 'encoding' features



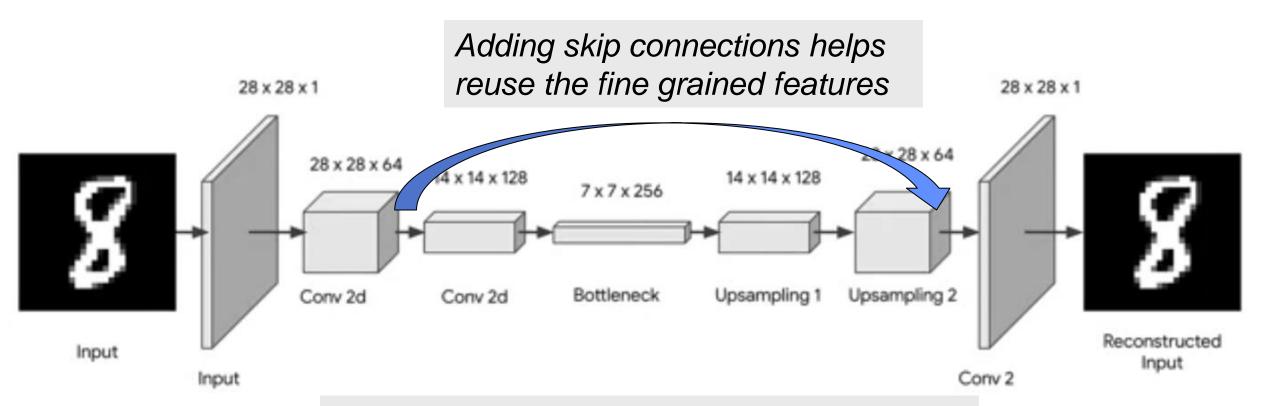
## A CNN architecture for MNIST autoencoding

more feature maps & downsampling : What if we want to 'encoding' features reconstruction the input? 28 x 28 x 64 14 x 14 x 128 7 x 7 x 256 Conv 2d Bottleneck Conv 2d Reconstructed Input Input Input

## A CNN architecture for MNIST autoencoding



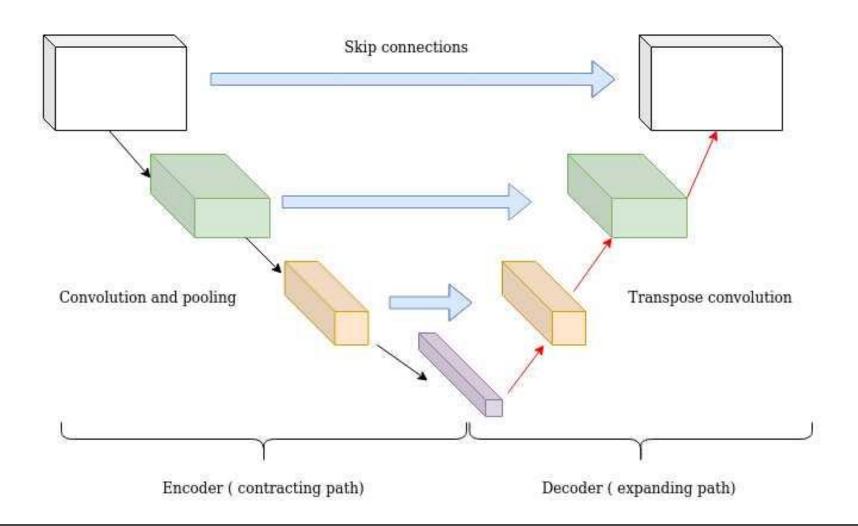
## A CNN architecture for MNIST autoencoding



NOTE the 28x28 encoded maps have to be skipped ahead to where the 28x28 decoding maps are – which axis is concatenated?



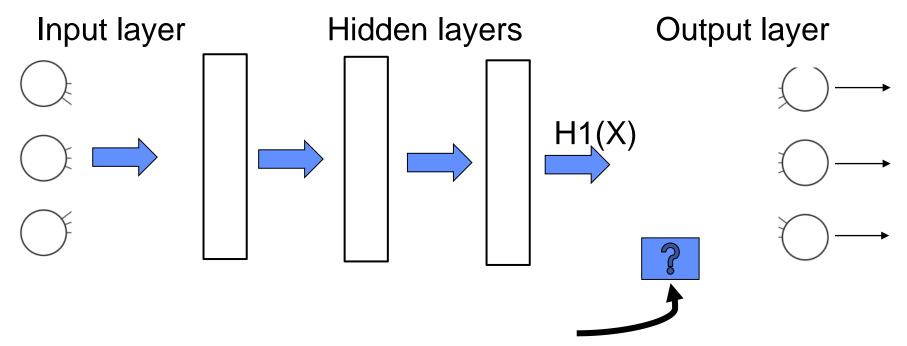
### Image Encoder-Decoder is a "UNET" architecture



#### **Outline**

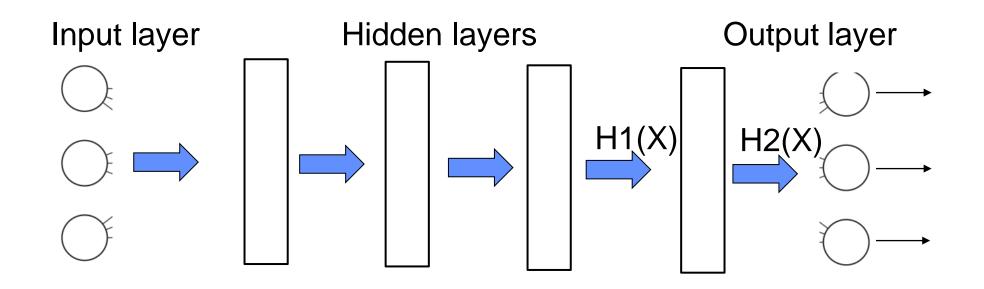
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### Consider: Can we keep adding deep layers?



Given some deep network, should I add another layer? What should a new layer learn?

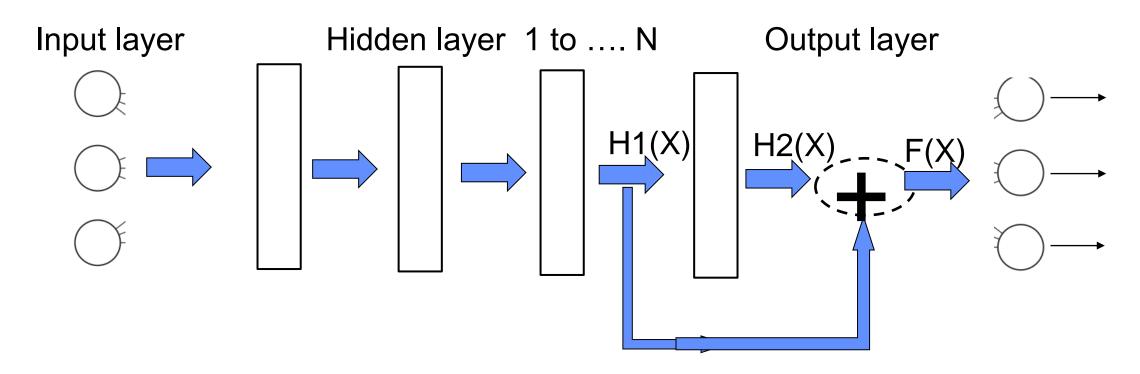
### Consider: Can we keep adding deep layers?



If H(X) is good then this new layer could be unnecessary,

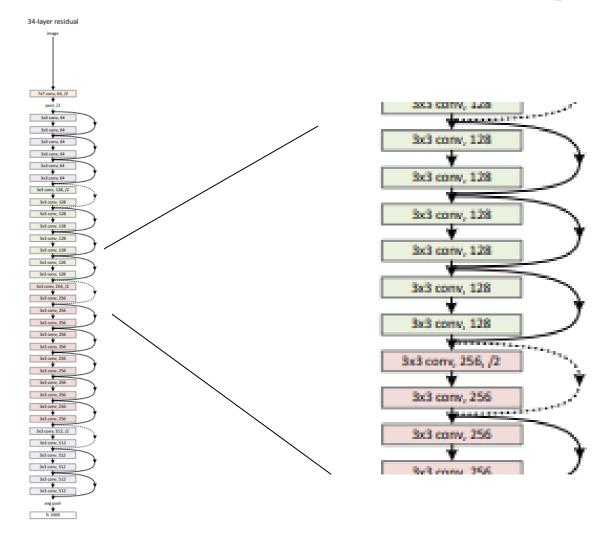
Eg H2(X) should be just H(X)

### Skip with addition makes a 'residual' connection



Make it easy for next layer to learn nothing – conversely, use F(X)=H2(X)+x so that H2(X)=F(X)-X. The H2() function learned is a residual function

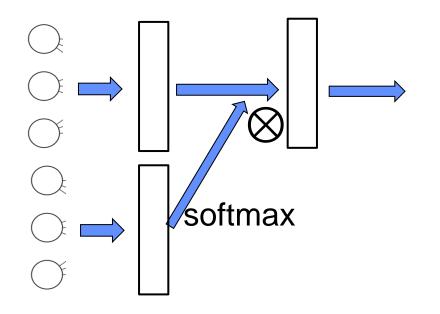
### "Resnet" residual connections help deeper learning



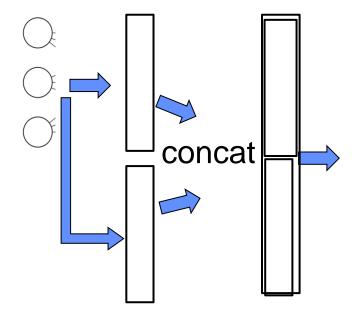
Deep Residual Learning, He et.al, 2015

## Summary: useful connections for architectures, and the intuitions

Softmax for gating

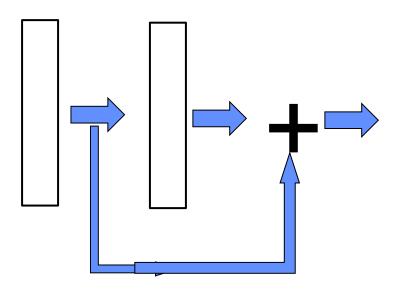


Skip connections for feature reuse



UNET, also feedforward nets..

Residual connections help deeper learning



Resnet, large image classification

Recurrent nets, language transformer nets



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### Keras: Sequential API VS Functional API

A sequence of layers: the inputs are assumed to be in order

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A sequence of layers: the inputs are assumed to be in order

A sequence of functions:
Input layer(s) are specified

### Keras: Sequential API VS Functional API

```
#specify the neural network model and learning parameters
my_model = tf.keras.models.Sequential([
                   tf.keras.layers.Flatten(input_shape=(28, 28)),
                   tf.keras.layers.Dense(32,activation='relu'),
                   tf.keras.layers.Dense(10,activation='softmax')])
mv model.summarv()
```

A sequence of layers: the inputs are assumed to be in order

A sequence of functions: Input layer(s) are specified #specify the neural network model and learning parameters

```
inputs
                    = tf.keras.layers.Input(shape=(28, 28, 1,))
inputs_flattened
                   = tf.keras.layers.Flatten()(inputs)
hidden layer
                    = tf.keras.layers.Dense(32,activation='relu')(inputs_flattened)
output_layer
                    = tf.keras.layers.Dense(10,activation='softmax')(hidden_layer)
my_model = tf.keras.Model(inputs,output_layer)
my_model.summary()
```

The Model() function figures out the full path(s) to connect the input(s) to output(s)



#### **Outline**

- Gate connection idea
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#### **Exercise**

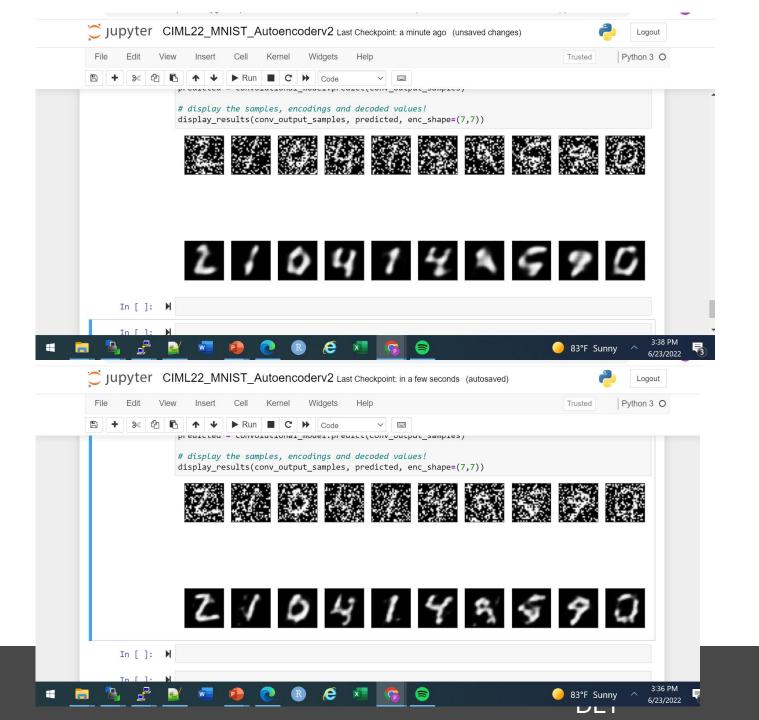
- MNIST autoencoder, reconstruct digits from noisy inputs
- Add skip connections with concatenation

Note: make sure you see how the outputs from encoding layers are matched up to inputs for decoding layers!

14x14 encoding feature maps should be concatenated with 14x14 decoding maps

Review outputs to see improvements





With out skip 20 epochs Loss 0.1664

With skip, 20 epochs loss 0.14

Are the numbers a little bit more reconstructed?