#### Capsule Networks

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#### Introduction

#### Digit Classification Task





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#### Digit Classification Task





#### Feature Learning

• Use Feed-forward neural networks.



#### Feature Learning

- Use Feed-forward neural networks.
- Use Convolutional neural networks.





#### Feature Learning

- Use Feed-forward neural networks.
- Use Convolutional neural networks.
- Use Capsule networks.



#### Convolutional Neural Network

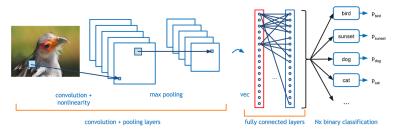
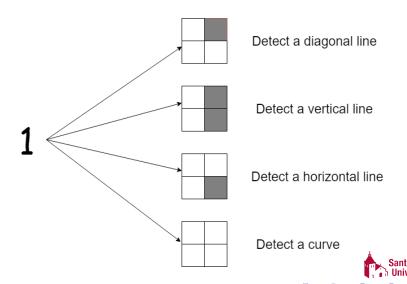


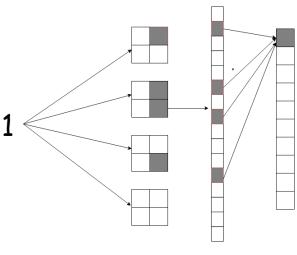
Figure: An example of a convolutional neural network<sup>1</sup>



#### Convolutional Layer

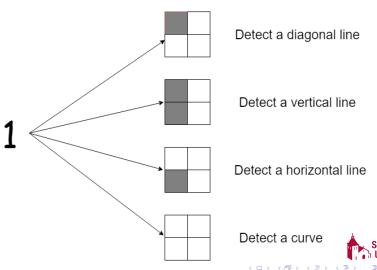


## Convolutional Layer



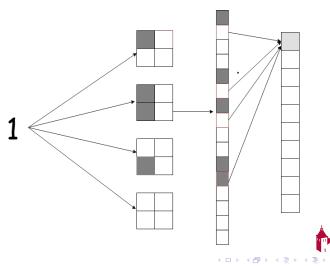
#### Max Pooling Layer

What if digit 1 is shifted to the left?



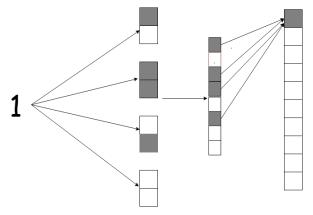
#### Max Pooling Layer

What if digit 1 is shifted to the left?



# Max Pooling Layer

Sub-sample the feature map.

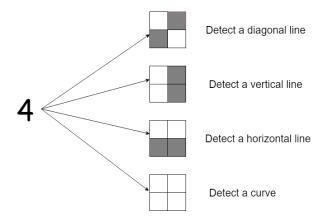


CapsNets



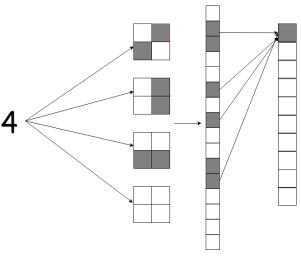
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# Try to recognize digit 4





# Try to recognize digit 4



#### What is the limitation of CNNs?

Why does CNNs poorly distinguish digit 4 from digit 1?



#### Capsule Networks

- Model a part-whole relationship.
- Each capsule represents one unique feature.
- A group of low-level capsules represents "part" of the object.
- A high-level capsule represents an object.
- Model a hierarchical structure of visual features.



# Try to recognize a face using CNNs

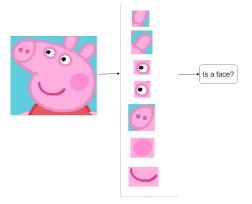


Figure: Detecting Peppa pig's face<sup>2</sup>



# Try to recognize a face using CNNs

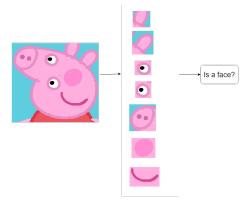
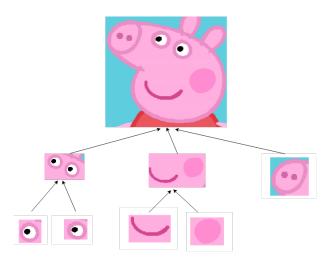


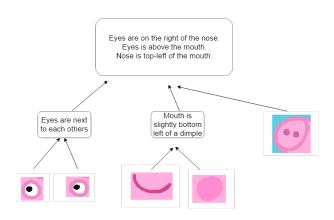
Figure: Detecting an incorrect Peppa pig's face



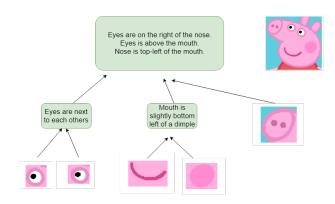
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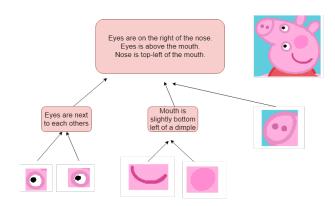
















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#### Capsule Approach

- Try to understand 3D space.
- Capsule encapsulates all important information about the state of the features they are detecting in a vector form.
- A vector length is a probability of detecting a feature.
- A vector direction is the state of the detected feature.

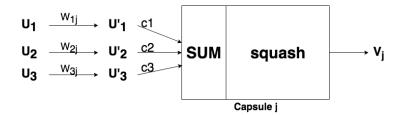




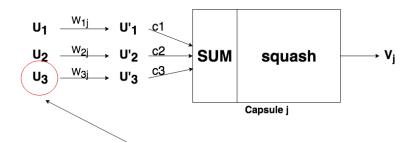
#### Capsule Approach

- Show a Capsule v.s. Traditional Neural.
- Draw a capsule layer.
- Draw a one-layer feedforward neural.



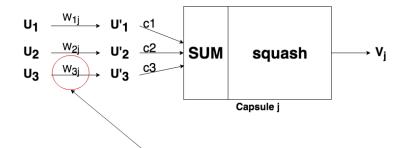






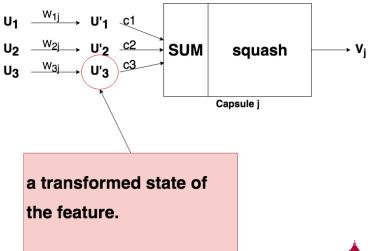
a vector that encodes the state of the feature.

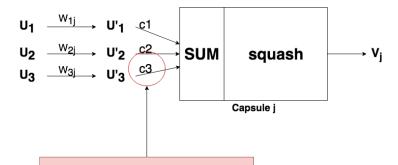




a learnable weight matrix that encodes spatial relationship

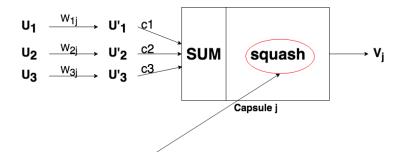






a non-negative scalar weight determined by routing algorithm.





a non-linear function that takes a vector as an input and outputs another vector.



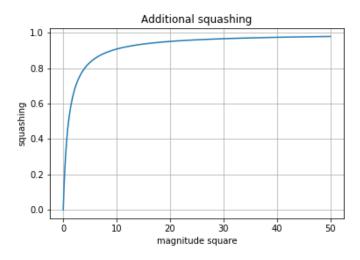
#### Squash Function

$$extbf{v}_j = rac{\| extbf{s}_j\|^2}{1+\| extbf{s}_j\|^2} rac{ extbf{s}_j}{\| extbf{s}_j\|}$$





#### **Squash Function**



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#### **Dynamic Routing**

- Each low-level capsule i, it has a weight  $c_{i,j}$  as a probability of its output belong to each high level capsule j.
- $\sum_i c_{i,j} = 1$
- $\forall i,j$   $c_{i,j} >= 0$





#### Dynamic Routing

- for all capsule i in layer I and capsule j in layer I + 1:  $b_{ii} = 0$
- a for r iterations do:
  - for all capsule i in layer  $l: c_i = \operatorname{softmax}(b_i)$
  - ② for all capsule j in layer  $l+1: s_i = \sum_i c_{ii} \hat{u}_{i|i}$
  - for all capsule j in layer  $l+1: v_i = \text{squash}(s_i)$
  - of for all capsule i in layer l and capsule j in layer l+1:  $b_{ii}=b_{ii}+\hat{u}_{i|i}\cdot v_i$
- return *v<sub>i</sub>*





#### Architecture for MNIST

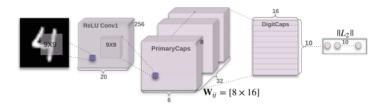


Figure: An encoder architecture<sup>3</sup>



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#### Architecture for MNIST

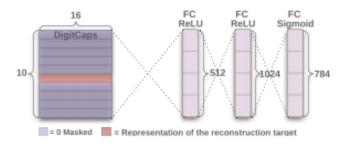


Figure: A decoder architecture<sup>4</sup>



#### Loss function

$$L_c = T_c \max(0, m^+ - \|V_c\|)^2 + \lambda(1 - T_c) \max(0, \|V_c\| - m^-)^2$$



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#### References

- https://medium.com/ai%C2%B3-theory-practice-business/ understanding-hintons-capsule-networks-part-i-intuition-b
- Oynamic Routing Between Capsules https://arxiv.org/abs/1710.09829



#### Questions

# Questions?



