



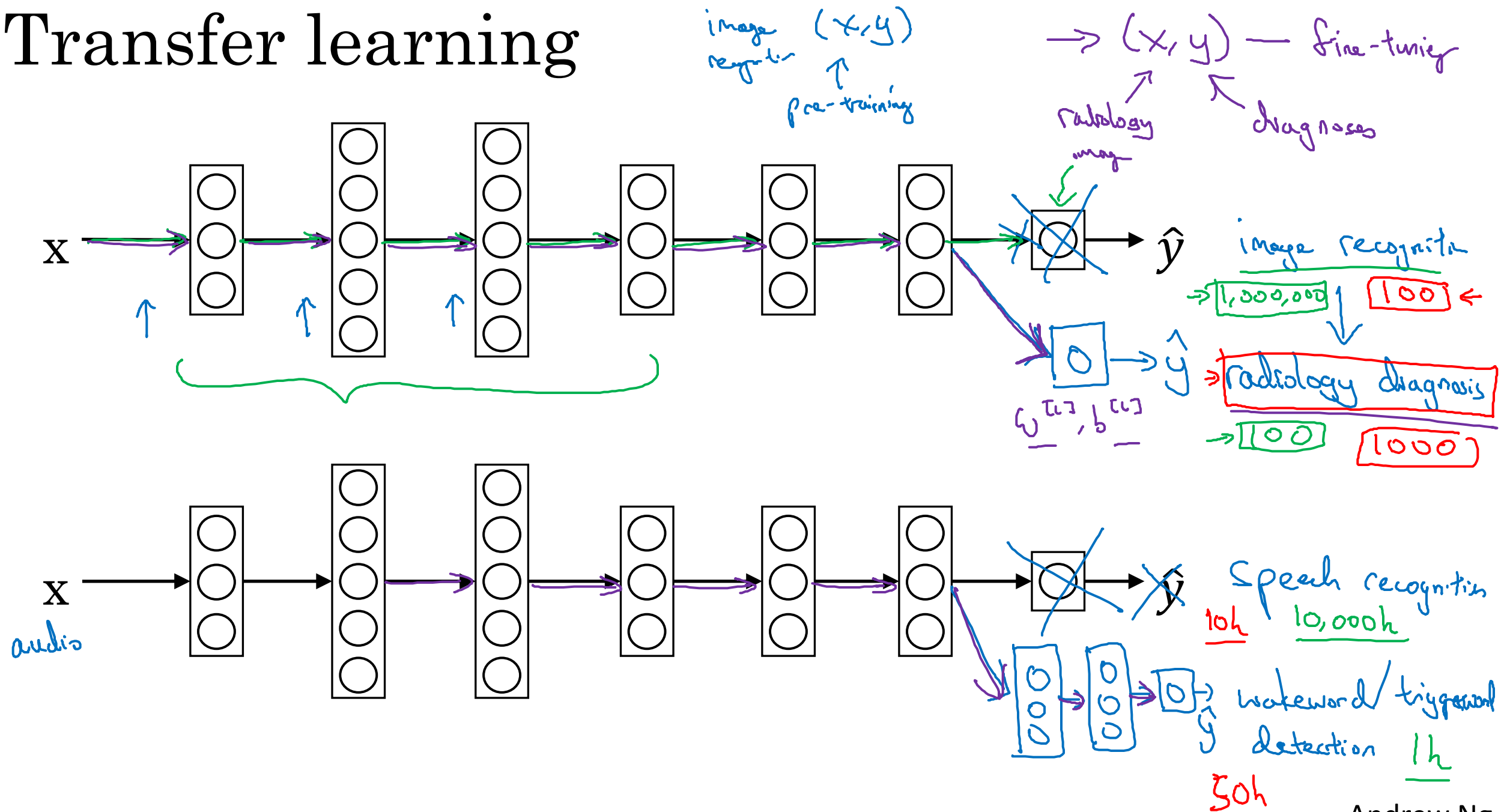
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Learning from  
multiple tasks

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Transfer learning

# Transfer learning



# When transfer learning makes sense

Transfer from A  $\rightarrow$  B

- Task A and B have the same input  $x$ .
- You have a lot more data for Task A than Task B.  

- Low level features from A could be helpful for learning B.



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Learning from  
multiple tasks

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Multi-task  
learning

# Simplified autonomous driving example



$x^{(i)}$

Pedestrians

Cars

Stop signs

Traffic lights

⋮

$y^{(i)}$

0

1

1

0

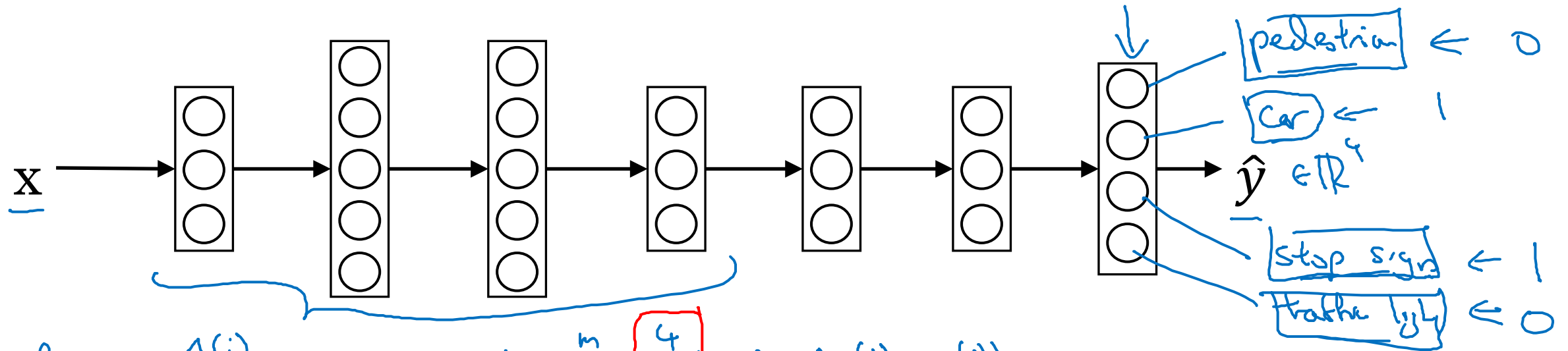
⋮

$(4, 1)$

$$Y = \begin{bmatrix} y^{(1)} & y^{(2)} & y^{(3)} & \dots & y^{(m)} \\ 1 & 1 & 1 & \dots & 1 \end{bmatrix}$$

$(4, m)$

# Neural network architecture



Loss:  $\frac{1}{m} \sum_{i=1}^m \sum_{j=1}^4 \mathcal{L}(\hat{y}_j^{(i)}, y_j^{(i)})$

Sum only over  
value of  $j$  with  
0/1 label.

Unlike softmax regression:  
One image can have multiple labels

Usual logistic loss

$$-y_j^{(i)} \log \hat{y}_j^{(i)} - (1 - y_j^{(i)}) \log (1 - \hat{y}_j^{(i)})$$

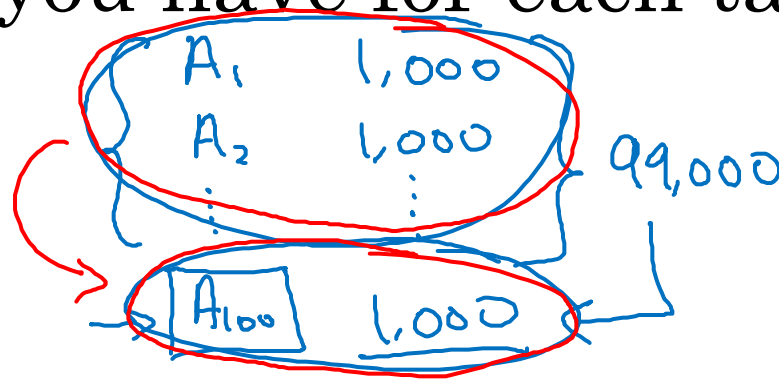
Multi-task learning  $\leftarrow$

$$Y = \begin{bmatrix} 1 & 1 & \dots & 1 & ? \\ 0 & 1 & \dots & 1 & ? \\ ? & ? & \dots & 1 & ? \\ ? & ? & \dots & 0 & ? \end{bmatrix}$$

# When multi-task learning makes sense

- Training on a set of tasks that could benefit from having shared lower-level features.
- Usually: Amount of data you have for each task is quite similar.

A     1,000,000  
↓     ↓  
B     1,000



- Can train a big enough neural network to do well on all the tasks.



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# End-to-end deep learning

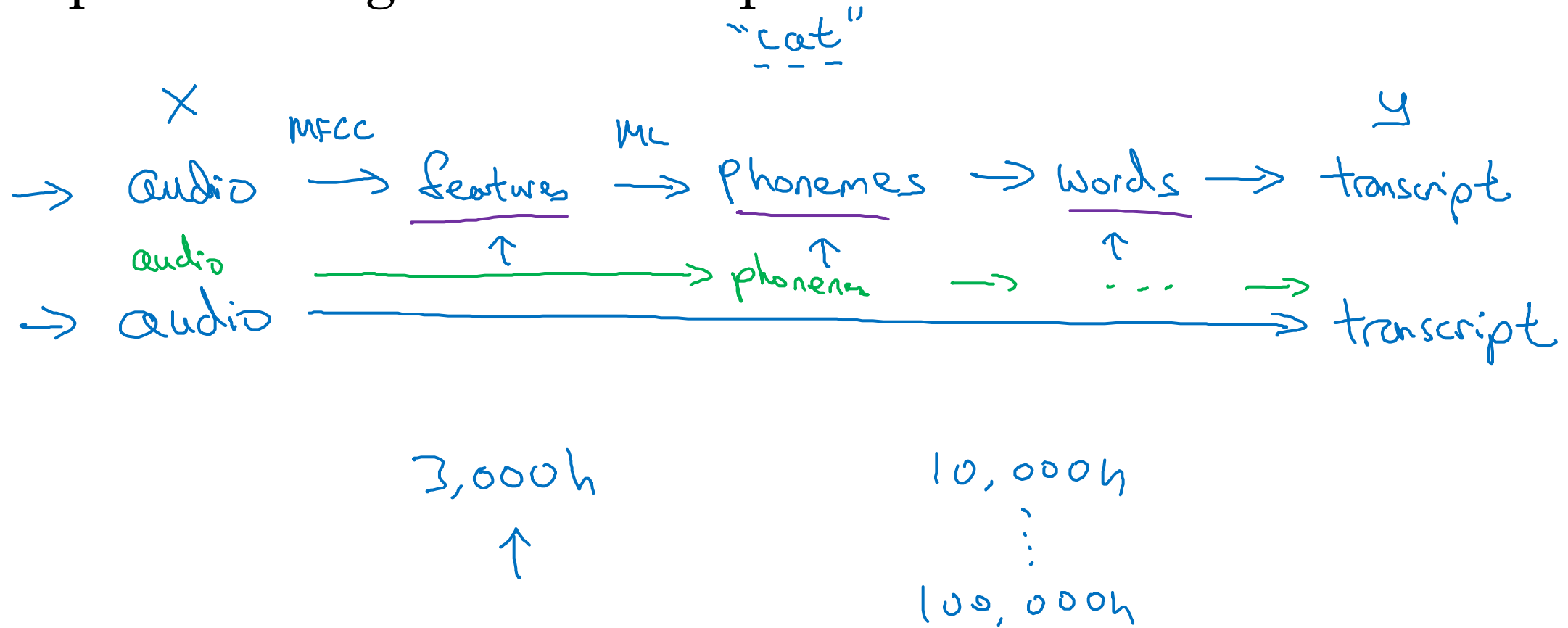
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## What is end-to-end deep learning



# What is end-to-end learning?

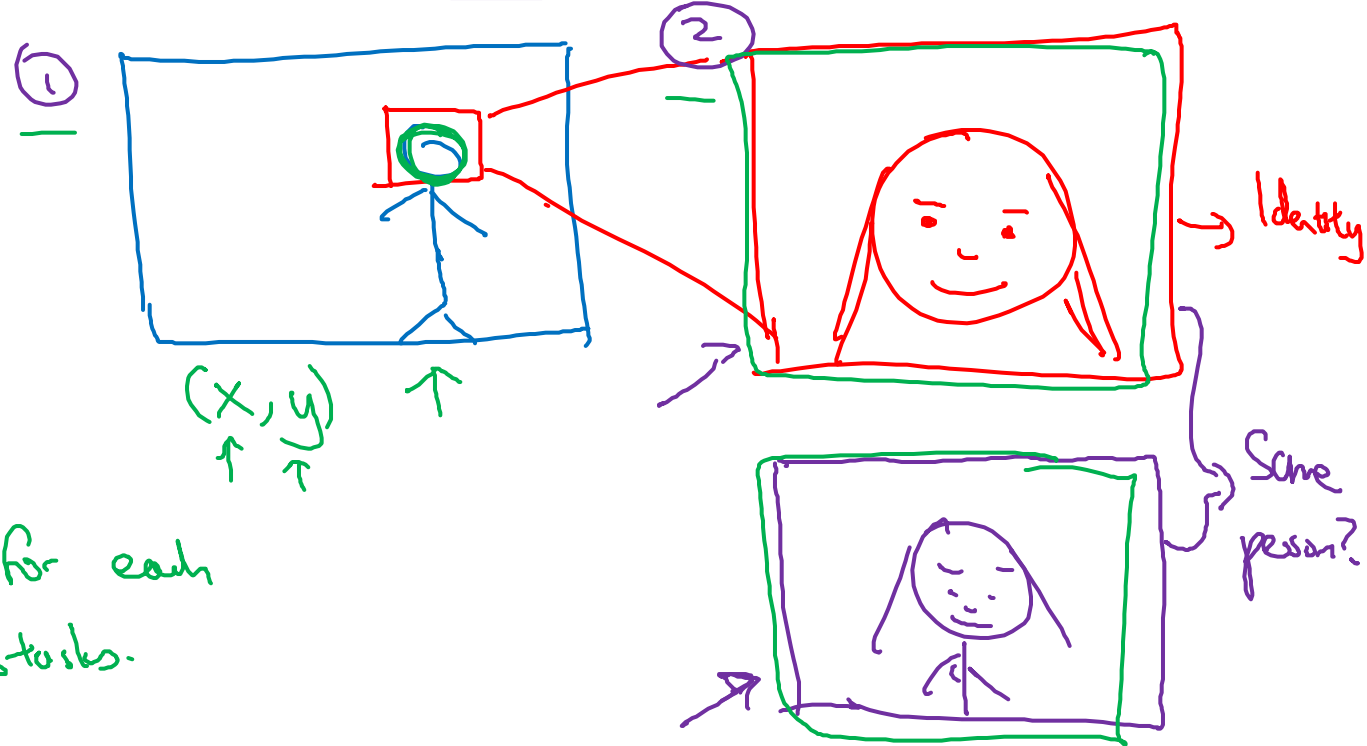
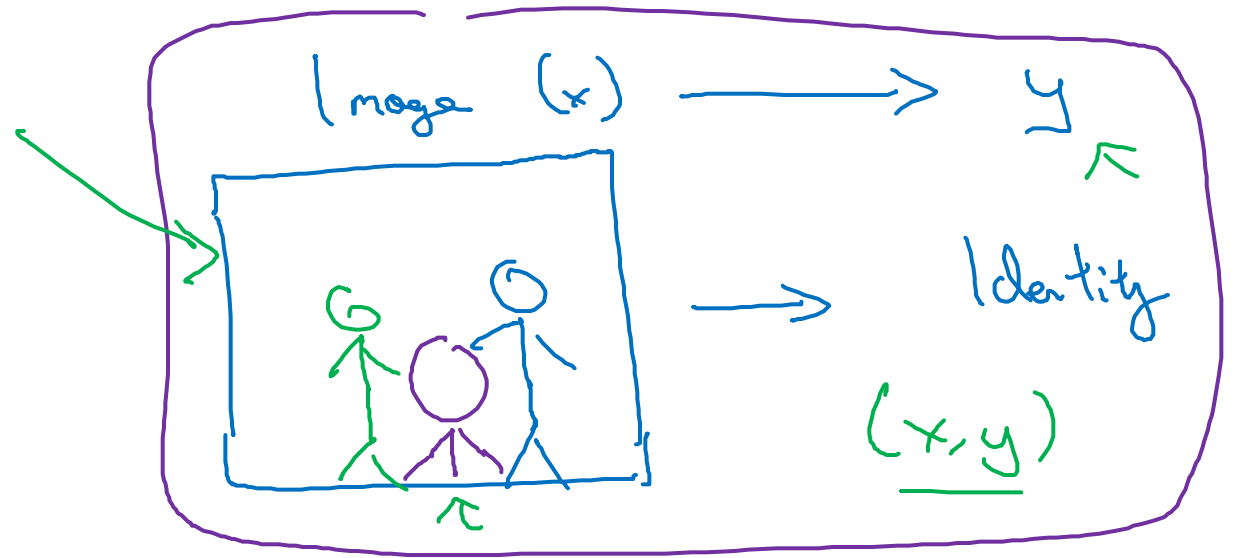
## Speech recognition example



# Face recognition



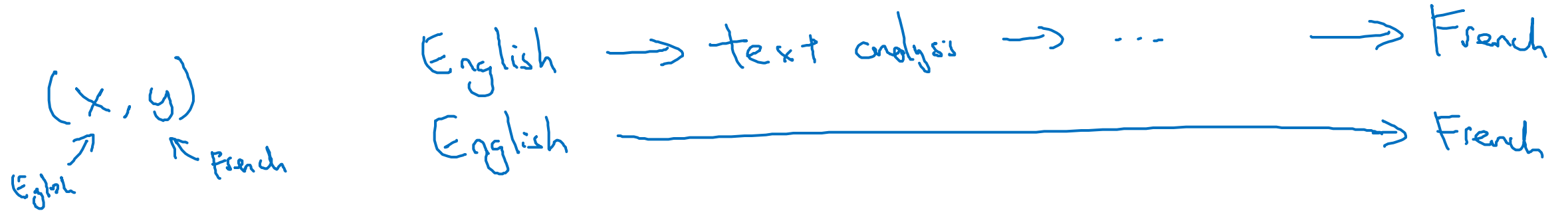
[Image courtesy of Baidu]



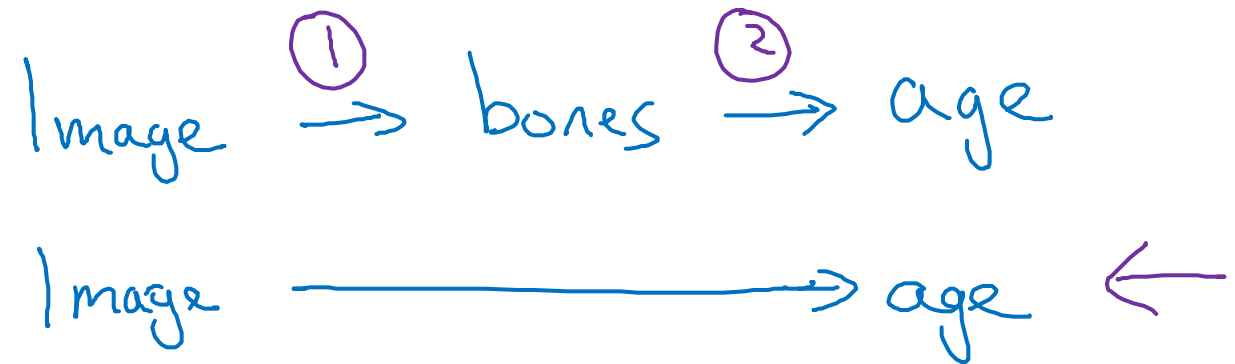
Have data for each  
of 2 sub-tasks.

# More examples

## Machine translation



## Estimating child's age:





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End-to-end deep  
learning

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Whether to use  
end-to-end learning

# Pros and cons of end-to-end deep learning

## Pros:

- Let the data speak
- Less hand-designing of components needed

$x \rightarrow y$

→ "phonemes"  
c a t

## Cons:

- May need large amount of data
- Excludes potentially useful hand-designed components

$x - - - - - \rightarrow y$

input  
end  
↓  
 $x \rightarrow y$   
output  
end  
↓

(x, y)

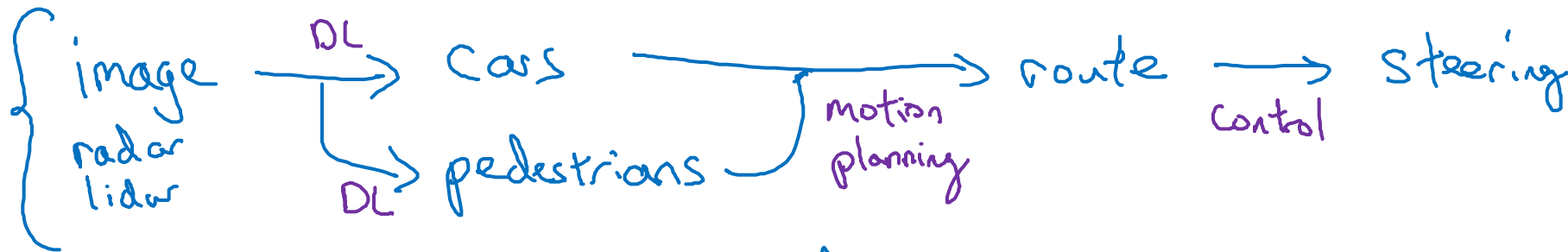
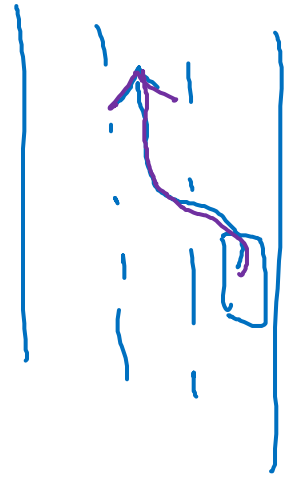
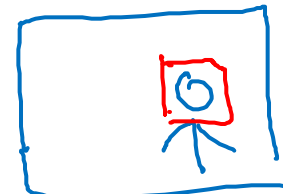
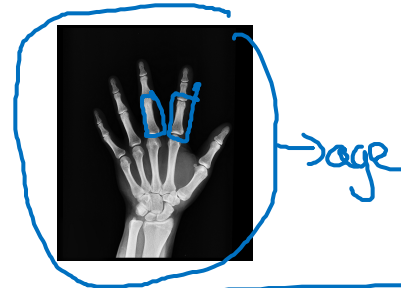
Data.  
- - -

Hand-design.  
- - -

# Applying end-to-end deep learning

Key question: Do you have sufficient data to learn a function of the complexity needed to map  $x$  to  $y$ ?

$x \rightarrow y$



- Use DL to learn individual components
- Carefully choose  $x \rightarrow y$  depending what tasks you can get data for.

$\rightarrow$  image  $\rightarrow$  steering