



deeplearning.ai

Mismatched training  
and dev/test data

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Training and testing  
on different  
distributions

# Cat app example

Data from webpages



core about this  
Data from mobile app

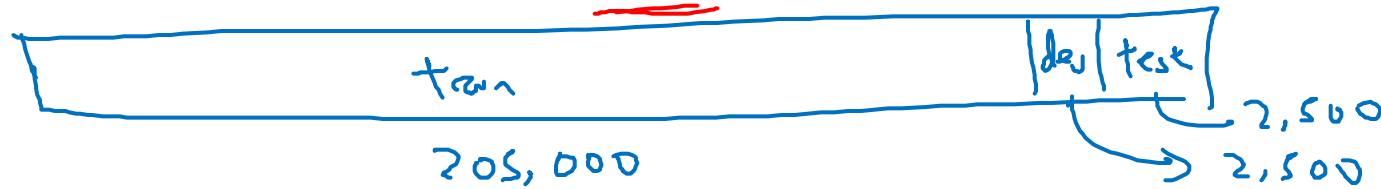


→ ≈ 200,000

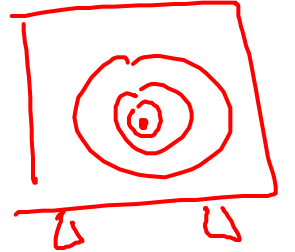
→ 210,000  
↓ shuffle

→ ≈ 10,000

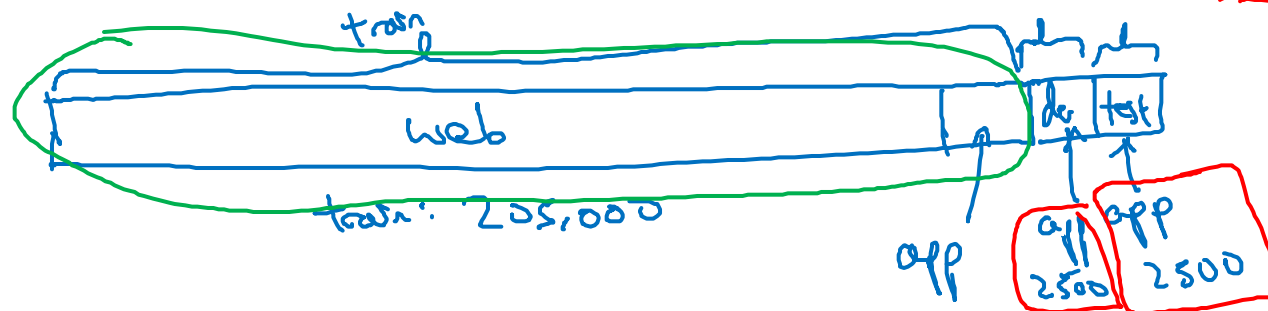
~~Option 1:~~



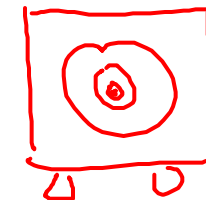
$\frac{200K}{210K}$



Option 2:



2381 - web  
119 - mobile app



# Speech recognition example

Speech activated rearview mirror



## Training

Purchased data

$\downarrow \downarrow$   
 $X, y$

Smart speaker control

Voice keyboard

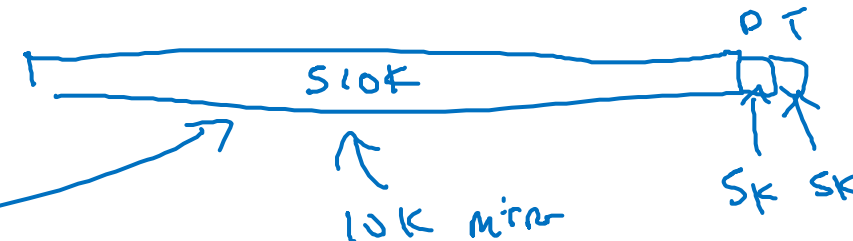
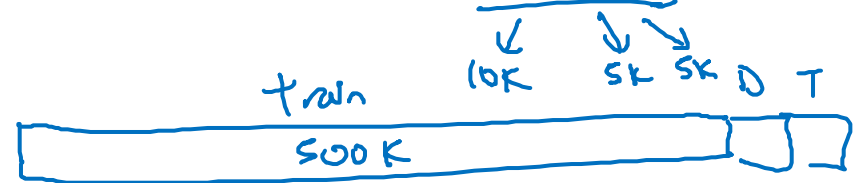
...

500,000 utterances

## Dev/test

Speech activated  
rearview mirror

→ 20,000





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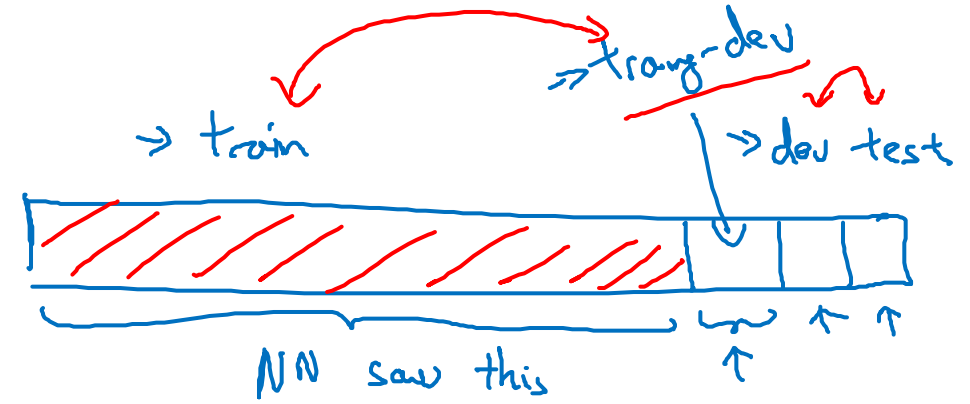
Bias and Variance with  
mismatched data  
distributions

# Cat classifier example

Assume humans get  $\approx 0\%$  error.

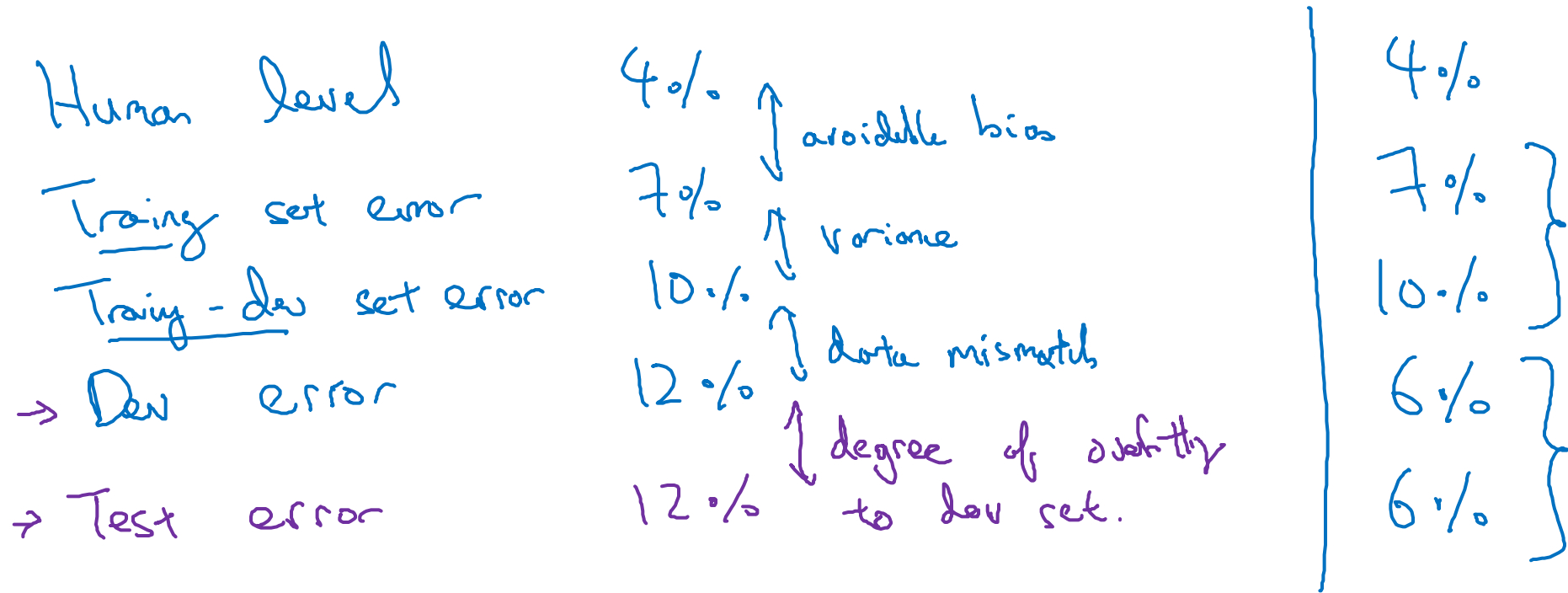
Training error .....  $1\%$   
 Dev error .....  $10\%$   $\downarrow 9\%$

Training-dev set: Same distribution as training set, but not used for training



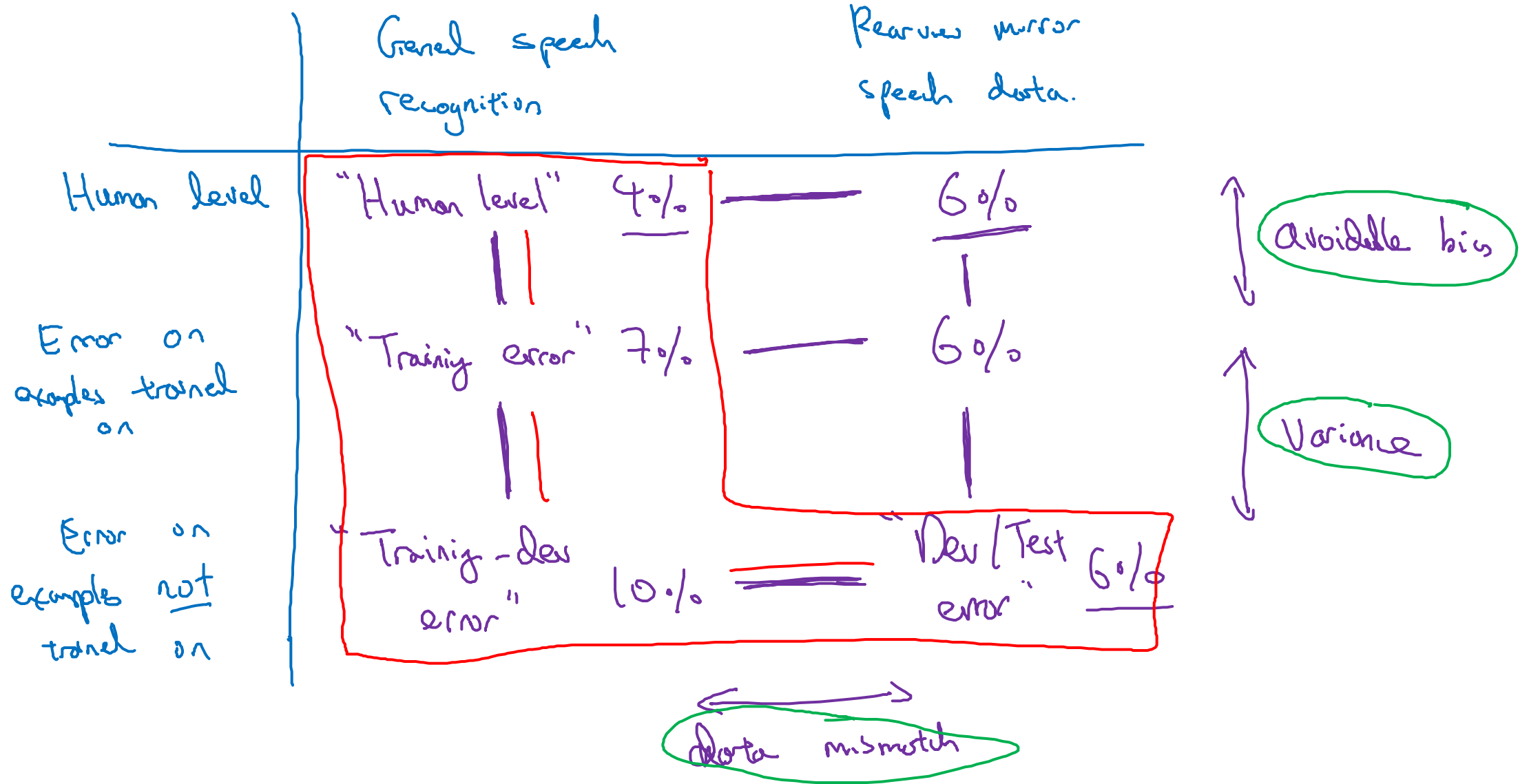
Training error	$1\%$	$\uparrow$ Variance	$1\%$	$\uparrow$ Variance
→ Training-dev error	$9\%$		$1.5\%$	$\uparrow$ Data mismatch
→ Dev error	$10\%$		$10\%$	
		Variance		
Human error - - -	$0\%$	$\uparrow$ Avoidable bias	$10\%$	$\uparrow$ Avoidable bias
Training error	$10\%$	$\downarrow$ bias	$10\%$	$\downarrow$ Variance
Training-dev error	$11\%$		$11\%$	$\uparrow$ Data mismatch
Dev error	$12\%$		$20\%$	
	Bias		Bias + Data mismatch	

# Bias/variance on mismatched training and dev/test sets



# More general formulation

Recurrent mirror





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Addressing data  
mismatch



# Addressing data mismatch

- • Carry out manual error analysis to try to understand difference between training and dev/test sets

E.g. noisy - car noise

street numbers

- • Make training data more similar; or collect more data similar to dev/test sets

E.g. Simulate noisy in-car data

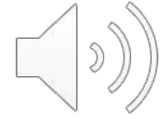
# Artificial data synthesis



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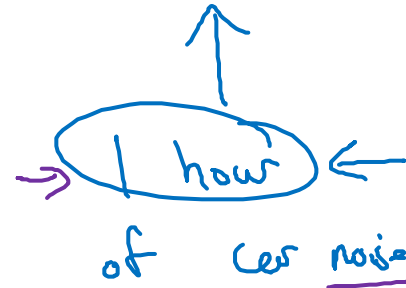
=



“The quick brown  
fox jumps  
over the lazy dog.”

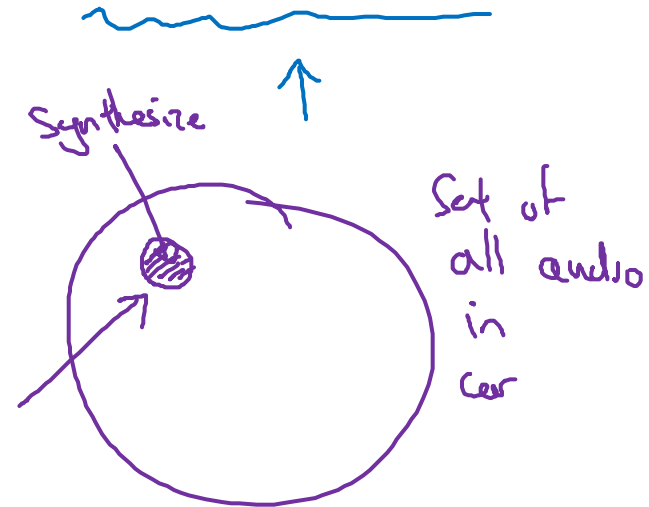
↑  
10,000 hours

Car noise



Overfit to 1 hour of  
car noise  
↑  
10,000 hours

Synthesized  
in-car audio



# Artificial data synthesis

Car recognition:



$\approx 20$  cars

