



deeplearning.ai

Setting up  
your goal

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When to change  
dev/test sets and  
metrics

# Cat dataset examples

$$\text{Error} : \frac{1}{M_{\text{dev}}} \sum_{i=1}^{M_{\text{dev}}} \underbrace{\mathbb{I}\{\hat{y}^{(i)} \neq y^{(i)}\}}_{0/1}$$

If you have put your goal  
↳ either metric has been wrongly defined  
↳ OR dev set data distribution is incorrect  
then you need to change your goal

Cat classifier

Metric: classification error

Algorithm A: 3% error

Algorithm B: 5% error

App goes through  
Internal DB of Images  
& spits out cat Images  
to show to users

"A" seems better, but say Algo A, in addition  
to spitting out cat Images, also classifies  
some naked women as cats & shows it  
to the user base  $\Rightarrow$  Bad for the company  
- But say "B" doesn't ever show porn Images  
 $\Rightarrow$  B is a Better Algo

Now, we can say our metric for Algo selection  
is wrongly defined, we blindly choose lower  
classification error, but "B" is better for production  
usage  $\Rightarrow$  New Error Metric

$$\frac{1}{M_{\text{dev}}} \sum_{i=1}^{M_{\text{dev}}} \left( w^{(i)} \mathbb{I}\{\hat{y}^{(i)} \neq y^{(i)}\} \right)$$

where

$$w^{(i)} = \begin{cases} 1 & \text{if } x^{(i)} \text{ is non-porn} \\ 100 & \text{if } x^{(i)} \text{ is porn} \end{cases}$$

# Orthogonalization for cat pictures: anti-porn

1. So far we've only discussed how to define a metric to evaluate classifiers. *(Place the target)*
2. Worry separately about how to do well on this metric. *(Aim & shoot at the target)*



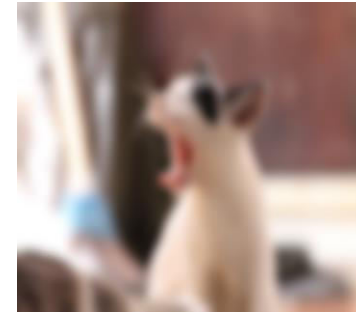
# Another example

Algorithm A: 3% error

Algorithm B: 5% error

(B is worse than A)  
objectively

Dev/test



However in the real world  
B performs better than A (say)  
- then you need to change your  
metric to pick your model  
+  
change dev/set to include  
blurry images

User images

If doing well on your metric + dev/test set does not correspond to doing well on your application, change your metric and/or dev/test set.