Your grade: 100%

Your latest: 100% • Your highest: 100% • To pass you need at least 80%. We keep your highest score

Next item \Rightarrow

1/1 point

1. Problem Statement

 $This \ example \ is \ adapted \ from \ a \ real \ production \ application, but \ with \ details \ disguised \ to \ protect$ confidentiality.



You are a famous researcher in the City of Peacetopia. The people of Peacetopia have a common characteristic: they are afraid of birds. To save them, you have to build an algorithm that will detect any bird flying over Peacetopia and alert the population.

The City Council gives you a dataset of 10,000,000 images of the sky above Peacetopia, taken from the city's security cameras. They are labeled:

- y = 0: There is no bird on the image
- y = 1: There is a bird on the image

Your goal is to build an algorithm able to classify new images taken by security cameras from Peacetopia.

There are a lot of decisions to make:

- What is the evaluation metric?
- How do you structure your data into train/dev/test sets?

The City Council tells you the following that they want an algorithm that $% \left(1\right) =\left(1\right) \left(1\right)$

- 1. Has high accuracy.
- 2. Runs quickly and takes only a short time to classify a new image.
- $3. \ \ Can \ fit \ in \ a \ small \ amount \ of \ memory, so \ that \ it \ can \ run \ in \ a \ small \ processor \ that \ the \ city \ will \ attach \ to$ many different security cameras.

 $You\ are\ delighted\ because\ this\ list\ of\ criteria\ will\ speed\ development\ and\ provide\ guidance\ on\ how\ to$ evaluate two different algorithms. True/False?

True:

False





\bigcirc Correct

Yes. More than one metric expands the choices and tradeoffs you have to decide for each with unknown effects on the other two.

2. After further discussions, the city narrows down its criteria to:

1/1 point

- "We need an algorithm that can let us know a bird is flying over Peacetopia as accurately as possible."
- "We want the trained model to take no more than 10 sec to classify a new image."
- "We want the model to fit in 10MB of memory."

If you had the three following models, which one would you choose?

Test Accuracy	Runtime	Memory size
98%	9 sec	9MB

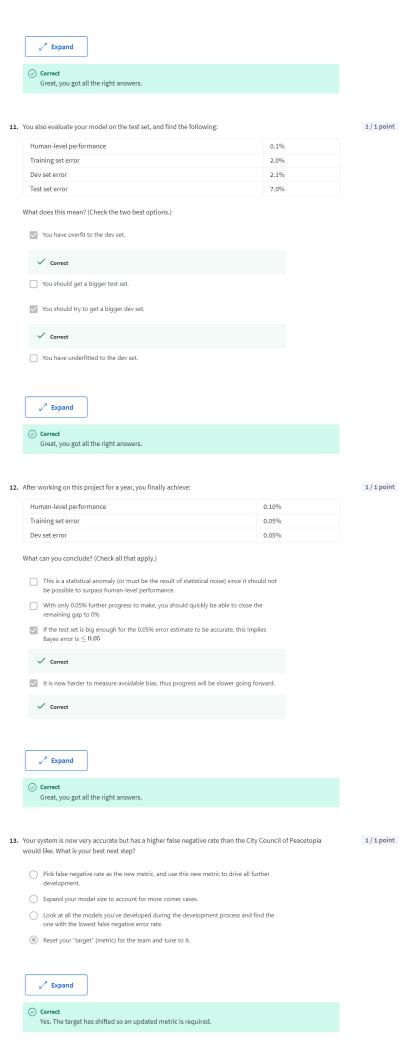
\circ	Test Accuracy	Runtime	Memory size
	99%	13 sec	9MB

0	Test Accuracy	Runtime	Memory size
	97%	1 sec	3MB

\bigcirc	Test Accuracy	Runtime	Memory size
	97%	3 sec	2MB

∠ ^A Expand	
Correct Correct! This model has the highest test accuracy, the prominent criteria you are looking for, compared with other models, and also has a runtime <10 seconds and memory size < 10MB.	
 The essential difference between an optimizing metric and satisficing metrics is the priority assigned by the stakeholders. True/False? False True 	1/1 point
✓ Correct Yes. Satisficing metrics have thresholds for measurement and an optimizing metric is unbounded. With 10,000,000 data points, what is the best option for train/dev/test splits?	1/1 point
 train - 95%, dev - 2.5%, test - 2.5% train - 33.3%, dev - 33.3%, test - 33.3% train - 60%, dev - 10%, test - 30% 	
Correct Yes. The size of the data set allows for bias and variance evaluation with smaller data sets. Now that you've set up your train/dev/test sets, the City Council comes across another 1,000,000 images from social media and offers them to you. These images are different from the distribution of images the City Council had originally given you, but you think it could help your algorithm. You should add the citizens' data to the training set. True/False? False True	1/1 point
 ✓ Correct Yes. This will cause the training and dev/test set distributions to become different, however as long as dev/test distributions are the same you are aiming at the same target. 	
 One member of the City Council knows a little about machine learning and thinks you should add the 1,000,000 citizens' data images to the dev set. You object because: (Choose all that apply) A bigger test set will slow down the speed of iterating because of the computational expense of evaluating models on the test set. The 1,000,000 citizens' data images do not have a consistent x→y mapping as the rest of the data. This would cause the dev and test set distributions to become different. This is a bad idea because you're not aiming where you want to hit. ✓ Correct Yes. Adding a different distribution to the dev set will skew bias. The dev set no longer reflects the distribution of data (security cameras) you most care about. ✓ Correct Yes. The performance of the model should be evaluated on the same distribution of 	1/1 point
images it will see in production.	

Great, you got all the right answers.		
. You train a system, and its errors are as follows (error = 100%-Accuracy)):	1/1 poin
		-/
Training set error Dev set error	4.0%	
Dev set error	4.5%	
This suggests that one good avenue for improving performance is to tra	nin a bigger network so as to drive	
down the 4.0% training error. Do you agree?		
Yes, because this shows your bias is higher than your variance.		
No, because this shows your variance is higher than your bias.		
No, because this shows your variance is higher than your bias.		
Yes, because having a 4.0% training error shows you have a high bias		
No, because there is insufficient information to tell.		
No, because there is insufficient information to tell.		
√ ⁿ Expand		
2 Expand		
⊘ Correct		
You want to define what human-level performance is to the city council answer?	which of the following is the best	1/1 poin
The average performance of all their ornithologists (0.5%).		
2.2.252 p.2.2.3. and containings (0.579).		
The performance of their best ornithologist (0.3%).		
The average of regular citizens of Peacetopia (1.2%).		
The average of all the numbers above (0.66%).		
∠ ⁿ Expand		
2 Expand		
○ Correct		
Yes. The best human performance is closest to Bayes' error.		
Which of the below shows the optimal order of accuracy from worst to	best?	1/1 poin
The learning algorithm's performance -> human-level performance -	> Bayes error.	
The learning algorithm's performance -> Bayes error -> human-level	performance.	
Human-level performance -> Bayes error -> the learning algorithm's	performance.	
Human-level performance -> the learning algorithm's performance -	> Rayes error	
Trained teres performance is the realising algorithms performance	- bayes anon	
∠ ⁿ Expand		
 Correct Yes. A learning algorithm's performance can be better than huma 	n-level performance but it can neve	er
be better than Bayes error.	n teres personnance out it can neve	
. You find that a team of ornithologists debating and discussing an imag	e gets an even hetter 0.1%	1 / 1 poin
performance, so you define that as "human-level performance." After w		
end up with the following:		
Human-level performance	0.1%	
Training set error Dev set error	2.0%	
Based on the evidence you have, which two of the following four option (Check two options.)	ns seem the most promising to try?	
Train a bigger model to try to do better on the training set.		
✓ Correct		
Try decreasing regularization.		
ny decreasing regulatization.		
✓ Correct		
Try increasing regularization.		
Get a bigger training set to reduce variance.		



14. Over the last few months, a new species of bird has been slowly migrating into the area, so the performance of your system slowly degrades because your data is being tested on a new type of data. There are only 1.000

_	es of the new species. The city expects a better system from you within the next 3 months. Which of these ld you do first?	
0) Put the new species' images in training data to learn their features.	
0) Split them between dev and test and re-tune.	
0	Augment your data to increase the images of the new bird.	
0) Add pooling layers to downsample features to accommodate the new species.	
	∠ [™] Expand	
(Yes. A sufficient number of images is necessary to account for the new species.	
with:	? (Check all that agree.)	
	? (Check all that agree.) With the experience gained from the Bird detector you are confident to build a good Cat detector on the first try. You could consider a tradeoff where you use a subset of the cat data to find reasonable performance with reasonable iteration pacing.	
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