Quiz, 12 questions

1 point	
1.	the idea of having 4v4 convolutions vectorable?
wny is	the idea of having 1x1 convolutions reasonable?
	They act like L2 regularization, reducing overfitting by making weights smaller.
	They act like dimensionality reduction, removing unnecessary feature maps from previous layer.
	They accelerate training by making loss function more convex.
	They accelerate inference by replacing fully-connected layers with convolutional layers.
1 point	
2. How ca	n one reduce computational burden suffered by the deep convolutional neural networks?
	Use 1x1 convolutions to reduce number of feature maps.
	Use stacked 3x3 filters to reduce the number of parameters in feature maps.
	Use 3x3 filter decomposition into 1x3 and 3x1 filters to reduce the number of parameters in feature maps.
	Use Adam optimizer instead of vanilla SGD to accelerate learning.
1 point 3.	ne correct statements.
iviai K tr	Residual connections help back propagate errors in very deep networks, leading to better generalization and handling the vanishing gradient problem.

Convo	Batch Normalization can help in CNNS, because the spatial dimensionality reduction makes covering larger butional ifpatiumgsefagewisualizecognition
Quiz, 12 qu	lestions With stochastic depth, the network (expected) depth reduces during testing while maintaining the full depth at training time.
	DenseNets are harder to train because of their complicated architecture.
point	
4.	
Why do	deep learning methods outperform everyone else in computer vision in most tasks?
	Visual features are learned automatically and therefore focused on a specific task.
	Neural networks allow us to recover the nonlinear and complex dependencies.
	Deep learning methods can be applied to any data set, as opposed to the classical ones.
	Computer power has reached a level that allows you to solve optimization problems with a variety of parameters.
1	
1 point	
point 5.	
point 5.	all methods of dealing with overfitting.
point 5.	Adding recurrent layers
point 5.	
point 5.	Adding recurrent layers
point 5.	Adding recurrent layers  Small random turns
point 5.	Adding recurrent layers  Small random turns  Increasing the optimization step
point 5.	Adding recurrent layers  Small random turns  Increasing the optimization step  Increasing resolution of images
point 5.	Adding recurrent layers  Small random turns  Increasing the optimization step  Increasing resolution of images  Early learning stop

10101111

point	
սiz, 12 qu 6	estions
6. <b>Why ca</b> i	n part localization be useful for fine-grained recognition problems?
	Parts may have visual features extracted at their original resolution which helps focus on subtle appearance differences between them.
	It speeds up training of neural networks because they have to process little data.
	It allows focusing on differences associated with specific object parts which can be small relative to the whole image.
	Parts are the only way to solve fine-grained classification tasks.
1 point	
7. Which o	of the following are valid examples of image similarities?
	Color similarity (get objects of the same color)
	Caption similarity (get images with similar captions)
	Scene geometry similarity (geometrically similar scenes)
	Instance similarity (get this very object)
1 point	
8.	
ror a lo	cal semantic hash of 10101111, which would be the closest neighbours of bit distance equal to 1?
	10101100
	10111110
	10101011
	00101111

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Quiz, 12 q	uestions
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9.	
How to	combine advantages of k-means and LSH clustering into a unified indexing scheme?
	Cluster image descriptors using k-means, then quantize the very same descriptors and concatenate cluster index and LSH mask into a joint signature.
	Cluster image descriptors using k-means, then compute LSH codes for the difference of original points and cluster centers using LSH.
	Compute long LSH codes for the original images, then cluster these using k-means.
	Just use k-means and LSH separately and see what works best.
1 point 10. Why do	t o we need a preprocessing of the face image in the problem of face identification?
	To reduce the impact of the diversity of human pose, angle, scale.
	To account for different types of camera.
	To search for a person on the basis of photographs.
	To account for the variability of the appearance of a person (make up, haircut).
1 point 11.	
What p	parts are used in CNN cascade for keypoints regression task?
	Generator and discriminator.
	Initial (robust) and refinement models.
	Multi-task predictors for different keypoints.
	Predictors from different scales in pyramidal architecture.

Quiz, 12 questions		
point		
12.		
Which method is the main one in the identification problem?		
Training of the classifier, compare the classification results.		
Training descriptor, the comparison of distances between descriptors.		
Applications of finding similar individuals, a comparison of intersection results are similar.		
The prediction of attributes, comparison of the predicted attributes.		
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