✅ **Congratulations! You passed!**

Grade received 100%   To pass 80% or higher

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Continuous Evaluation and Monitoring

Total points 6

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1. Many ML models suffer from declining predicting capabilities over time. A common solution used to overcome this deterioration is to keep retraining your model with new data. As part of this process you may encounter a phenomenon called concept emergence. Which of the following statements accurately describes this emergent phenomenon?

   1 / 1 point

○ The persistent appearance of stationary data that remains immutable over time.

○ The lack of covariate shift.

◉ The appearance of new patterns in data distribution that were not previously present in your dataset.

○ The loss of prediction quality over time.

✓ **Correct**
That's right! These new patterns are the new concepts emerging in your data. Continuous evaluation and monitoring will help you address them properly.

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2. Statistical process control is a technique that detects concept drift assuming that the errors follow a binomial distribution. Would the system trigger an alarm if $p_t = \sigma_t = 0.3$ and $p_{min} = \sigma_{min} = 0.12$?

   1 / 1 point

◉ Yes

○ No

✓ **Correct**
That's right! The values provided satisfy the alarm rule: $p_t + \sigma_t \geq p_{min} + 3\sigma_{min}$

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3. In sequential analysis you detect concept drift by calculating the negative predictive value, precision, recall, and specificity of the system based on a standard contingency table. If the data is stationary these quantities should not change over time. This analysis is tedious as it requires recomputing all these metrics each time we get a new sample. Which of the following approaches is usually implemented to overcome this problem?

   1 / 1 point

○ Monte Carlo sampling

○ Recursive computation and caching

○ Adaptive windowing

◉ Incremental update rule

✓ **Correct**
That's right! The incremental update rule is $P_*^t \leftarrow \eta_* P_*^{t-1} + (1 - \eta_*) I_{y_i = \hat{y}_i}$

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4. Drift detection techniques in unsupervised settings typically suffer from the curse of dimensionality. Which of the following techniques is an appropriate solution to mitigate the effects of this curse? (Check all that apply)

   1 / 1 point

☐ SVD (Singular Value Decomposition)

☑ NMF (Non Negative Matrix Factorization)

✓ **Correct**
That's right! NMF is a very useful dimensionality reduction technique when the features are constrained to be all non-negative.

☐ K-means

☑ PCA (Principal components analysis)

✓ **Correct**
That's right! Principal components analysis is the right tool to reduce the number of features to detect drift more efficiently.

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5. In unsupervised settings, clustering is a very useful method to detect novelty in your data. In this method, you cluster the incoming batches of data to one of the known classes. If you observe that the features of the new data

   1 / 1 point

cluster the incoming batches of data to one of the known classes. If you observe that the features of the new data are lying far away from the classes of known features, you can term it as an emerging concept. The downside of this method is that it detects only _____ drift and not _____ changes.

○ feature, cluster-based

⦿ cluster-based, population-based.

○ cluster-based, feature

○ population-based, cluster-based

⊘ **Correct**
That's right! Drift is detected only on a cluster centric view,

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6. It is a sad truth that most of the machine learning models are trained with a fixed set of stationary data. It is very likely that in this process you may have slightly biased your model in favor of your limited data at training. Consequently, as time progresses, your ML model's performance will deteriorate with time. Monitoring helps prevent this performance decay in which ways? (Check all that apply)

1 / 1 point

☑ Reduces false alarm rates

⊘ **Correct**
Correct! If applied wisely, monitoring can detect data drift early on and adjust the model accordingly to adapt to these changes and hence improving model's performance.

☑ Allows you to identify distribution changes close to the classification boundaries

⊘ **Correct**
Yes! Data drift can change classification boundaries quite drastically and monitoring will help you detect and mitigate this unwanted behavior.

☑ Identify regions in latent space where the model performs poorly

⊘ **Correct**
That's right! Monitoring will help you identify areas in latent space where your model struggles at classification. You can further use this knowledge to refine your model.

☐ By performing dimensionality reduction

☐ Allows you to establish ground truth labels

☐ By retraining your model constantly