

# Machine learning exam

This exam is timed and the duration of each question is displayed at the end of each question. A progress bar below each question also indicates the time remaining.

**Questions are automatically skipped** as soon as the time runs out, and **there is no turning back**.

It is divided into 2 parts

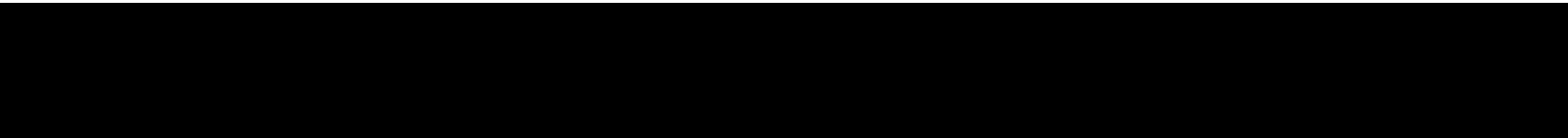
**1- Oral exam:** you will read the questions on the screen and answer them orally to your examiner.

You will have a two-minute break in between to prepare.

**2- The written exam:** you will read the questions and answers on the screen, and you will have to answer via a google forms form by ticking the right answer(s).

# **ORAL EXAM 15 min**

(2 min)



# QUESTION 1

Explain the difference between supervised and unsupervised machine learning. (1 min)

# QUESTION 2

What is the F1-score metric, and when is it preferred?

(1 min 30)



# QUESTION 3

What are the hyperparameters of a decision tree? (4 at least) (1 min)

# QUESTION 4

Compare cross-entropy and mean squared error (MSE) as cost functions. (1 min)

# QUESTION 5

What is the model selection problem? (1 min)



# QUESTION 6

Provide examples of approaches to deal with highly imbalanced datasets. (1 min 30)

# QUESTION 7

Why is it important to split a dataset into training and test sets in a supervised ML task? (1 min)



# QUESTION 8

What is feature engineering, and why is it important?

(1 min)



# QUESTION 9

How does the Random Forest algorithm improve over a single decision tree? (1 min)

# QUESTION 10

Increasing the number of features in a dataset always improves a model's performance.

True or False? Explain why if false. (*1 min*)

# QUESTION 11

What is KNN, and in what context is it used (1 min)

# QUESTION 12

Name three types of supervised machine learning algorithms. (1 min)

# QUESTION 13

How can overfitting be avoided? (1 min)

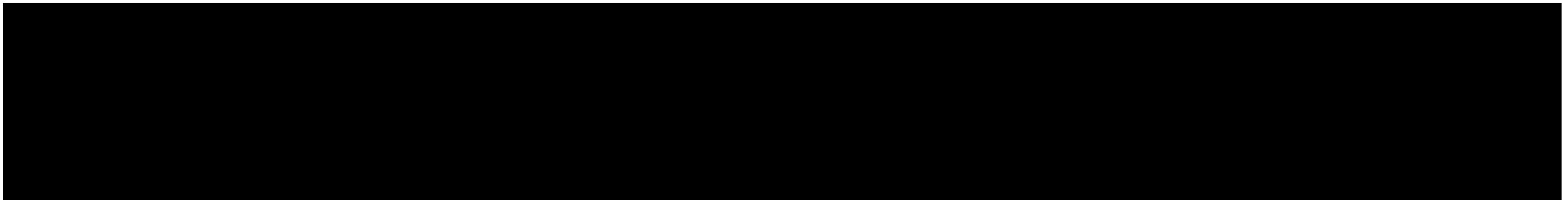


# QUESTION 14

Explain the difference between linear regression and logistic regression (1 min)

# **WRITTEN EXAM 15 min**

(2 min)



# QUESTION 1

Name 3 metrics that can be used to evaluate the output of a classification model? *(1 min)*

# QUESTION 2

What is the formula for accuracy? *(1 min)*



# QUESTION 3

What is the Recall value of the following confusion matrix. Provide the result in the form of a simplified fraction. (2 min)

The formula in the image defines Recall (True Positive Rate) as follows:

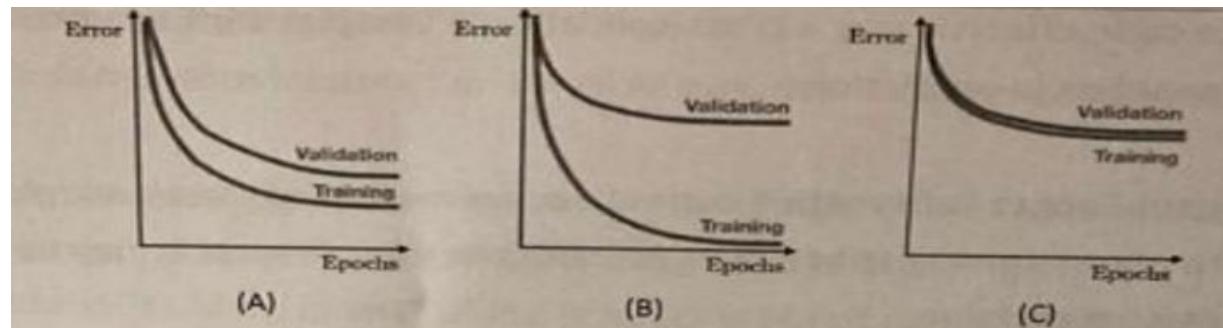
		Actual Values	
		Positive (1)	Negative (0)
Predicted Values	Positive (1)	50	10
	Negative (0)	5	35

$$\text{Recall} = \frac{TP}{TP + FN}$$

# QUESTION 4

**Match each curve to the corresponding proposition :** (1min30)

- a) (A) Good Fit, (B) Underfit, (C) Overfit
- b) (A) Good Fit, (B) Overfit, (C) Underfit
- c) (A) Underfit, (B) Overfit, (C) Good Fit
- d) (A) Underfit, (B) Good Fit, (C) Overfit



# QUESTION 5

**To measure the effectiveness of a split while constructing a Decision Tree,  
we can use:** (1 min)

- a) The Classification error metric
- b) The Gini index metric
- c) The entropy metric
- d) None of the above

# QUESTION 6

**Which technique can be used to handle missing data in a dataset? (1 min)**

- a) Removing columns with missing values.
- b) Imputation using the mean or median.
- c) Using predictive models to estimate the missing values.
- d) All of the above.

# QUESTION 7

**Which of the following scenarios can be considered an unsupervised machine learning problem? (1 min)**

- a) Predicting the price of a house.
- b) Identifying groups of customers with similar behaviors.
- c) Determining whether an email is spam or not.
- d) Predicting future sales of a product.

# QUESTION 8

**What does bagging stand for? (1 min)**

- a) Boosted Aggregation
- b) Bootstrap Aggregation
- c) Bayesian Aggregation
- d) Balanced Aggregation

# QUESTION 9

**Choose the correct answer(s):** (1 min)

- a) The more over-fitted a model is, the larger its bias-error is.
- b) The more over-fitted a model is, the larger its variance-error is
- c) Regularization is a technique to reduce the bias-error
- d) Regularization is a technique to reduce the variance-error.

# QUESTION 10

**What is a limitation of Grid Search?** (1 min 30)

- a) It cannot tune hyperparameters for regression models.
- b) It requires large computational resources for high-dimensional grids.
- c) It only works for tree-based models.
- d) It cannot handle categorical hyperparameters.

# QUESTION 11

Why do we calculate both MSE and R2 score and what is the difference between these metrics? (1 min)

```
# 1. Load the dataset
data = pd.DataFrame({
    "X1": np.random.rand(100),
    "X2": np.random.rand(100),
    "X3": np.random.rand(100),
    "y": np.random.rand(100) * 10
})
# 2. Define features and target
X = data[["X1", "X2", "X3"]]
y = data["y"]
# 3. Split the dataset
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# 4. Initialize and train the Lasso model
lasso_model = Lasso(alpha=0.1)
lasso_model.fit(X_train, y_train)
# 5. Make predictions
y_pred = lasso_model.predict(X_test)
# 6. Evaluate the model
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
# 7. Display results
print("Mean Squared Error:", mse)
print("R2 Score:", r2)
print("Lasso Coefficients:", lasso_model.coef_)
```

# QUESTION 12

What does grid\_search.best\_params\_ return? (2 min)

```
# Split the dataset
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# 3. Define the SVM model
svm_model = SVC()
# Define the parameter grid for Grid Search
param_grid = {
    'C': [0.1, 1, 10, 100], # Regularization parameter
    'kernel': ['linear', 'rbf', 'poly'], # Kernel type
    'gamma': ['scale', 'auto'] # Kernel coefficient
}
# Perform Grid Search
grid_search = GridSearchCV(estimator=svm_model, param_grid=param_grid, cv=5, scoring='accuracy')
grid_search.fit(X_train, y_train)

best_params = grid_search.best_params_
best_model = grid_search.best_estimator_
# Make predictions
y_pred = best_model.predict(X_test)
# Evaluate the model
accuracy = accuracy_score(y_test, y_pred)
```

# **END OF EXAM**

(2 min)

