

MAD Quizz

Exercise 1 Mixture

Questions	Réponses
1. Is a mixture of uniform distributions identifiable ?	<input type="checkbox"/> yes <input checked="" type="checkbox"/> no
2. The kmeans algorithm assumes that	<input checked="" type="checkbox"/> the proportions of the mixture are all equal <input type="checkbox"/> the mean vectors of the mixture are equal <input checked="" type="checkbox"/> the covariance matrices of the mixture are equal <input checked="" type="checkbox"/> the covariance matrices are proportional to the identity <input checked="" type="checkbox"/> the components of the mixture are normal
3. The E-step of the EM algorithm computes	<input checked="" type="checkbox"/> The expectation of the complete log-likelihood <input type="checkbox"/> The probabilities $P(Z_{ik} = 0 \mathbf{x}_i)$ <input type="checkbox"/> The probabilities $P(\mathbf{x}_i Z_{ik})$
4. The M-step of the EM algorithm maximises	<input checked="" type="checkbox"/> the expectation of the log likelihood knowing X: $\log P(X)$ <input type="checkbox"/> the expectation of the log likelihood knowing Z: $\log P(Z)$ <input type="checkbox"/> something else
5. The EM algoirthm finds a maximum which is	<input type="checkbox"/> global <input checked="" type="checkbox"/> local
6. A finite mixture is	<input type="checkbox"/> gaussian <input checked="" type="checkbox"/> a density
7. The Bayesian information criterion	<input checked="" type="checkbox"/> is used for model selection <input type="checkbox"/> is used for determining the entropy of the mixture

Exercise 2 Principal Component Analysis

Questions	Réponses
1. Principal Component Analysis is used to identify ?	<ul style="list-style-type: none"> <input checked="" type="checkbox"/> new independent variables <input type="checkbox"/> clusters
2. The principal components are	<ul style="list-style-type: none"> <input type="checkbox"/> vectors spanning a subspace where data is projected <input checked="" type="checkbox"/> a new coordinates system <input type="checkbox"/> a gaussian mixture
3. The total variance of the dataset is equal to	<ul style="list-style-type: none"> <input type="checkbox"/> the maximum eigenvalue of the empirical covariance matrix <input checked="" type="checkbox"/> the sum of the eigenvalue of the empirical covariance matrix <input checked="" type="checkbox"/> the number of variables if the data is scaled.
4. The principal axis are	<ul style="list-style-type: none"> <input checked="" type="checkbox"/> vectors spanning a subspace where data is projected <input checked="" type="checkbox"/> eigenvectors of the empirical covariance matrix <input checked="" type="checkbox"/> orthogonal vectors.
5. The percentage of projected variance	<ul style="list-style-type: none"> <input checked="" type="checkbox"/> is used for choosing the number of principal component <input type="checkbox"/> is used for choosing the minimal eigenvalue.
6. Principal Coordinate Analysis aims at	<ul style="list-style-type: none"> <input type="checkbox"/> maximizing the total variance <input type="checkbox"/> summarizing the principal components <input checked="" type="checkbox"/> finding a representation of the dataset.
7. Principal Component Analysis	<ul style="list-style-type: none"> <input checked="" type="checkbox"/> allows a decomposition of the total variance <input checked="" type="checkbox"/> maximizes the projected variance <input checked="" type="checkbox"/> minimizes the variance in the space orthogonal to the projection space.
8. Principal Component Analysis requires	<ul style="list-style-type: none"> <input checked="" type="checkbox"/> more individuals than variables <input checked="" type="checkbox"/> more variables than individuals.