

1.1. Main Points Covered in Week 7 and 8

Week 7 – Support Vector Machines (SVM):

- Introduction to SVM for linearly separable and non-separable data.
- Kernel trick and non-linear SVM for handling complex decision boundaries.
- Support Vector Regression (SVR) for regression tasks.
- Statistical learning theory of SVM (VC dimension, margin theory).
- Multi-class classification methods in SVM (e.g., one-vs-rest, one-vs-one).
- Practical implementation of SVM in Python using **linear, polynomial, and RBF kernels**.

Week 8 – KNN and Decision Trees:

- Overview of the K-Nearest Neighbors (KNN) algorithm and its variants.
- Theory of KNN and importance of choosing the best number of neighbors (K).
- Decision trees: structure, splitting rules, entropy, and information gain.
- Regression trees and classification trees.
- Algorithms for building decision trees (e.g., CART, ID3, C4.5).
- Model complexity, pruning techniques, and bias-variance trade-off.
- Advantages and disadvantages of decision trees.
- Practical implementation of **KNN and decision trees in Python**.

1.2. Summary of Reading List (External Resources, etc.)

- **Textbook chapters:** Machine Learning texts often cover SVM (kernels, optimization) and decision trees (splitting, pruning).
- **Websites:** Scikit-learn documentation (sklearn.svm, sklearn.neighbors, sklearn.tree) gives practical code examples.
- **External tutorials:** Towards Data Science, Analytics Vidhya, and Medium blogs provide intuitive explanations and visualization for SVM and decision trees.
- **Code libraries:**
 - **Scikit-learn** → SVC, SVR, KNeighborsClassifier, DecisionTreeClassifier, DecisionTreeRegressor.
 - **NumPy & Pandas** for handling datasets.

- **Matplotlib/Seaborn** for visualization of decision boundaries and model results.


1.3. Reflection on Knowledge Gained




From these two weeks, I gained deeper insights into **classification and regression algorithms**.

- SVM taught me how **maximizing the margin** improves generalization and how the **kernel trick** enables handling non-linear data. I also understood that SVM is powerful but can be computationally expensive with large datasets.
- With KNN, I learned its **simplicity and intuition** but also its sensitivity to the choice of **K** and computational inefficiency in large datasets.
- Decision trees showed me how machine learning models can be **interpretable**, but also prone to **overfitting** unless pruning is applied.
- Implementing these in Python reinforced the theory and showed me how **hyperparameters (kernel choice, depth, neighbors)** directly affect performance.


Overall, this week improved my understanding of the **balance between model complexity, interpretability, and generalization** in machine learning.




Quiz Submissions - Week 7 Quiz ▾

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Individual Attempts	Grade
Attempt 1	 7 / 10 - 70 %
Attempt 2	 9 / 10 - 90 %
Overall Grade (highest attempt):	 9 / 10 - 90 %

Quiz Submissions - Week 8 Quiz ▾

 Add to ePortfolio

Individual Attempts	Grade
Attempt 1	 8 / 10 - 80 %
Attempt 2	 10 / 10 - 100 %
Overall Grade (highest attempt):	 10 / 10 - 100 %