| Machine Learning | | Computer Vision | | NLP | | |
|---|---|-------------------------------------|---|---|---|--|
| Adult dataset | Bi Predict whether an individual's income exceeds \$50K annually | | Bird species recognition | | Spam email detection | |
| Objectives Considerations | Predict whether an individuals income exceeds \$50K annually based on demographic and employment data. Analyze factors that significantly contribute to income classification. Data preprocessing steps Model building (from a selection of how choices) & hyperparameter furning You can see some baseline model performance in the provided link. Make sure you can achieve at least these numbers. | Objectives Considerations | Identify species of birds from images with high precision. Understand which visual features (e.g., beak shape, wing patterns) are most effective for classification. Technical Advice: Data exploration Data augmentation | Objectives Considerations | Classify emails as spam or not spam with high accuracy, identify features or patterns that are most indicative of spam emails. Text preprocessing Featur engineering ML models: non DL | |
| | | | Consider multiple DL models, including transfer learning | | Any imbalanced data? | |
| Dry Bean Data set | | | | | | |
| Objectives | Classify dry beans into their respective types based on physical properties. | Car Model recogni | Car Model recognition | | sis | |
| | | | Classify car models from images, focusing on make, model, and | | Determine the sentiment (positive, negative, neutral) of text | |
| | Determine the most distinguishing features for bean classification. | Objectives | year. | Objectives | data. Explore how different aspects of language influence perceived | |
| Considerations | Data preprocessing steps Choice of classifiers, take note of its high dimensionality Techniques used to address the dimensionality | Considerations | Determine key visual cues that differentiate car models. Data exploration Several approaches to attention mechanisms, pros and cons Understand feature extraction w.r.t to your dataset | Considerations | sentiment. Use of TFIDF or word embeddings Models: non DL vs DL | |
| Rice data set | | | Onder stand leature extraction w.r.t to your dataset | POStagging | | |
| Objectives | Differentiate between Cammeo and Osmancikrice varieties using physical characteristics. | Airplane Model recognition | | Objectives | Automatically label words in text with their corresponding part of speech. Achieve high accuracy in tagging, especially for ambiguous | |
| Considerations | Identify key traits that most affect the classification. Data preprocessing steps | Objectives | Recognize specific aircraft models from images. Highlight features like wing design, tail shape, or engine type that are critical for classification. | Considerations | | |
| | Feature dimension reduction? | Considerations | Data exploration | | Use pre-trained models BERT to improve performance | |
| | Choice of Al algorithms - supervised learning vs CNN? (CNN: acc=77.5%) | | DL model for fine grain recognition - possible transfer learning? | | | |
| Wine quality | | Dog breed recognition | | Machine translation Objectives Translate text from one language to another with high fidelity. | | |
| Objectives | Predict wine quality based on physicochemical tests. | Objectives | Classify dog breeds from images. | Objectives | Handle nuances like idioms, cultural references, and context in translation. | |
| Considerations | Discover the relationships between wine components and perceived quality. Data preprocessing steps - skewed data? | Considerations | Identify which features like fur pattern, ear shape, or body proportions are most indicative of breed. Data augmentation | Consderations | Use Transformer models? Metrics BLEU, COMET | |
| | Models: Consider side effects of skewed data and how to interpret the results in actual use case | | Explore transfer learning Ensemble learning? | Word embedding | | |
| | | | | | Represent words as vectors in a continuous vector space where | |
| Bank marketing dataset Predict whether a client will subscribe to a term deposit following | | | | Objectives | semantic relationships are preserved. Enhance downstream NLP tasks by providing better word | |
| Objectives | a marketing campaign. | Scene recogntion | | | representations. Compare different embedding models from non-contextual to | |
| Considerations | Understand which marketing strategies are most effective. Data preprocessing steps | Objectives | Recognize indoor scenes or environments from images. Determine which scene elements (texture, layout, objects) are most crucial for scene classification. | Considerations | contextual Embedding dimensions? | |
| Considerations | Feature engineering e.g. time series features | Considerations | Data exploration: spatial and contextual challenges may include lighting conditions, viewpoints, clutter | | Fine tune for domain specific inputs? | |
| | How do you deal with imbalanced data? Models: Classifier for imbalanced dataset? | | differences etc. | Building a chatbo | ot | |
| | Through this exercise, can you figure out the relationship between certain features and campaign outcome? | Face based identity recognition | | Objectives | Develop a conversational AI capable of understanding and | |
| Abalone dataset | | Objectives | Identify individuals based on facial features. Explore the robustness of facial recognition under varying lighting and expression conditions. | Considerations | contexts or intents. Transformer models? | |
| Objectives | Estimate the age of abalone from physical measurements. Explore the relationship between physical characteristics and | Considerations | Considerations Checkout some existing models eg Facenet etc. | | Prompt engineering and controlled generation via temperature setting etc. | |
| | age. | | | | User interactions - advanced feature will be multi modal | |
| Considerations | Data preprocessing steps | Facial expression | | Others | | |
| | Relationship between Age and features abalone need not be linear | Objectives | Classify emotions from facial expressions. | TSP in NSW | Solve the Travelling Salesman Problem (TSP) for a set of | |
| | Outlier management | | Understand the nuances of emotion detection across different cultures or individuals. | Objectives | locations within New South Wales (NSW), aiming to minimize the total distance travelled while visiting each location exactly once and returning to the starting point. | |
| | Models: Choice of models and which strategy do you use? | Considerations | Focus on facial areas like eyes, mouth? | Considerations | Compare different algorithmic approaches to understand their efficiency and effectiveness on real-world geographic data. Search methods: Consider Simulated annealing, genetic algo, | |
| | Understand the metrics and the respective meaning. | Hand gesture reco | Hand gesture recognition to play game | | ant colony optimization Simpler approach: DP? | |
| | | Objectives | Recognize hand gestures for game control. Optimize gesture recognition for real-time, low-latency interaction. | Understand Alph | | |
| | | Considerations | Gesture segmentation? | Objectives | Gain insight into how AlphaGo uses AI, specifically deep learning and reinforcement learning, to master the game of Go. Understand the implications of AlphaGo's achievements for AI research and game theory. | |
| | | Style transfer for in Objectives | Transfer artistic styles from one image to another. Explore the balance between style and content preservation in | Considerations | Monte Carlo Tree Search? With NN vs without NN Limitations and strengths of AlphaGo | |
| | | Considerations | the generated images. Neural style transfer | | | |