

Using Multiresolution Learning for Transfer in Image Classification

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Abstract

Our work explores the transfer of knowledge at multiple levels of abstraction to improve learning. By exploiting the similarities between objects at various levels of detail, multiresolution learning can facilitate transfer between image classification tasks.

We extract features from images at multiple levels of resolution, then use these features to create models at different resolutions. Upon receiving a new task, the closest-matching stored model can be generalized (adapted to the appropriate resolution) and transferred to the new task.

Main Idea

- Related objects have features that appear similar at lower resolutions.
- Low-resolution knowledge can transfer to related objects.
- High-resolution knowledge helps to differentiate between objects.



The Single-Resolution Classifier Single Resolution Data Classifier $\rightarrow R_k(X)$ \rightarrow Prediction The standard machine learning classifier is equivalent to c_0 . **The Multiresolution Classifier** Multiple Classifier Data **Resolutions** $R_{k}(X)$ → Prediction Multiple resolutions are Res $R_{r-1}(X)$ extracted by the Multi-scale or Multi-band method **The Multiresolution Ensemble** Multiple Ensemble of Combine Resolutions Classifiers Predictions $R_{0}(X)$ $R_{r-1}(X)$ Res Predictions y_0, \dots, y_{r-1} are combined using a linear discriminant or SVM











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