

# Beyond 2D-Grids: A Dependence Maximization View on Image Browsing

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## Abstract

- Many existing image search engines arrange returned results in some default order on the screen, typically the **relevance to a query** (keyword), only.
- Arguably, a more flexible and intuitive way would be to sort images into **arbitrary structures** such as grids, hierarchies, or spheres so that images that are **visually or semantically alike are in proximal locations**.
- Arbitrary structures pose **challenges** as computing cross-similarities between images and structure coordinates will be a difficult task.
- We instead exploit a recently developed machine learning technique: **kernelized sorting**.
- We extend the technique so that some images can be **pre-selected to guide** the layouting process.

## Kernelized Sorting

Kernelized sorting (Quadrianto et al. 2009) is

- a general technique to perform matching between pairs of objects from different domains which only requires a similarity measure within each of the two domains.

Some advantages of using kernelized sorting for image browser are

- producing a non-overlapping layout to enhance user's ability to have a good global overview of massive amount of images;
- providing flexibility in choosing the underlying structure, can be 2D grids, spirals, spheres, or even hierarchical structures.

Algorithm

**Input** Two sets of objects  $X = \{x_1, \dots, x_m\}$  and  $Y = \{y_1, \dots, y_n\}$

Compute kernel similarity matrix  $K$  on set  $X$

Compute kernel similarity matrix  $L$  on set  $Y$

Center the kernel matrices:

$$\bar{K} := HKH \text{ and } \bar{L} := HLH \text{ with } H_{ij} = \delta_{ij} - m^{-1}$$

**while** not converge **do**

Solve linear assignment problem

$$\pi_{i+1} \leftarrow \operatorname{argmax}_{\pi \in P_m} [\operatorname{tr} \bar{K} \pi^T \bar{L} \pi_i]$$

$$\text{with } P_m := \left\{ \pi \in \mathbb{R}^{m \times m} \text{ where } \pi_{ij} \geq 0 \text{ and } \begin{cases} \sum_i \pi_{ij} = 1 \\ \sum_j \pi_{ij} = 1 \end{cases} \right\}$$

**end while**

**Return** Locally optimum permutation matrix  $\pi^*$

## Kernelized Sorting with Preference Constraint

Motivation

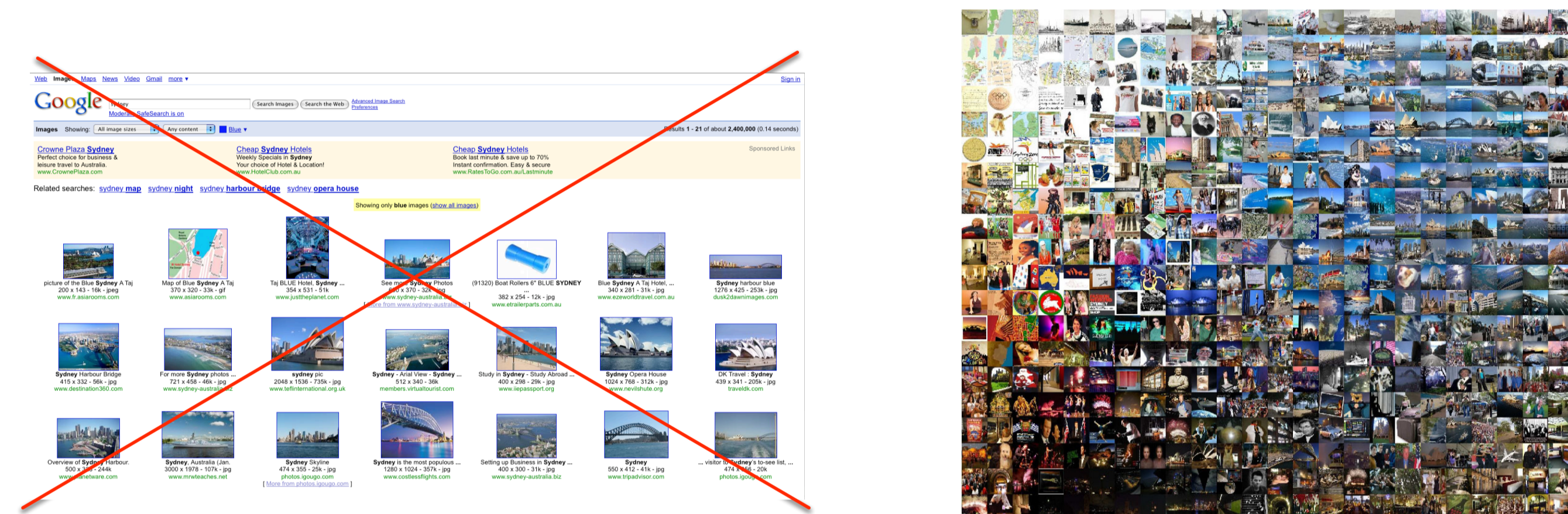
- Kernelized sorting technique does not accommodate injection of information or preferences to steer the layout of the images.
- We want to allow the user to express a small amount of preferences, for example blue images are placed at "north pole" of the sphere whereas black images are placed at the "south pole".

Solution

- We modify the constraint sets (while retaining the objective function un-touched) of the original kernelized sorting to express preferences.
- After some manipulations, we end up at the problem amenable to the same optimization technique as the original kernelized sorting.

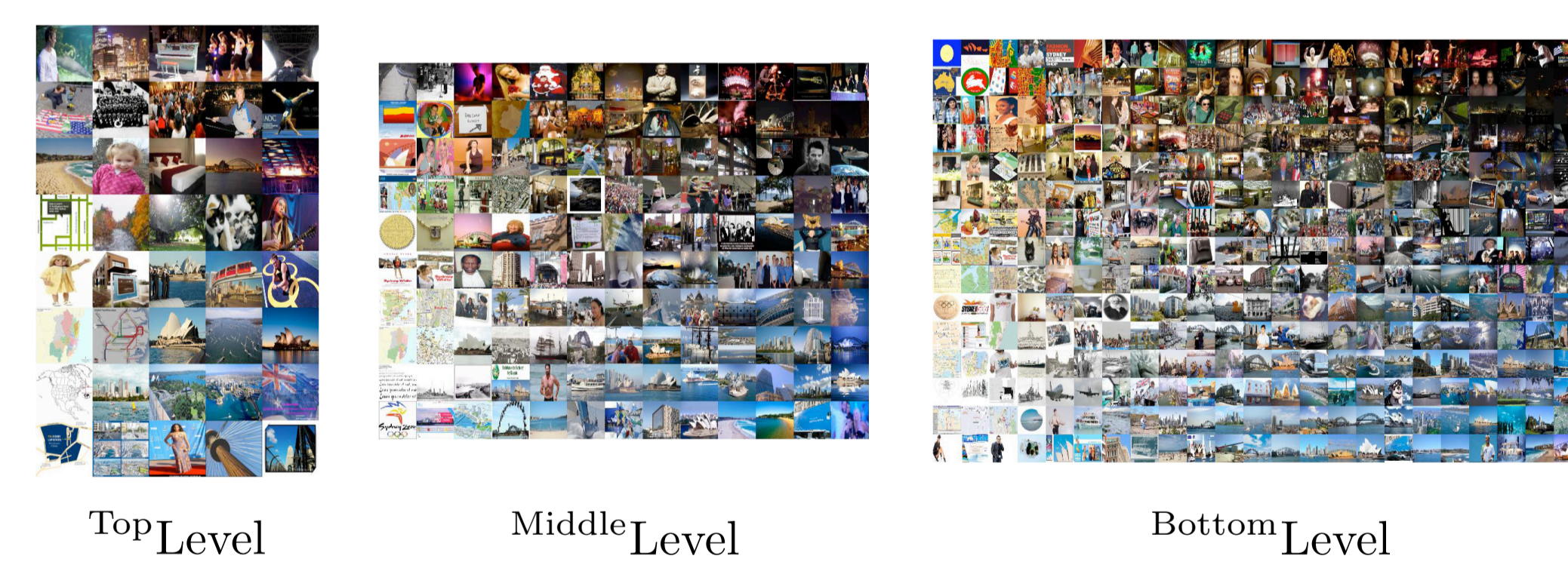
## Applications

### Bridging Keyword-based and Semantic-based Search

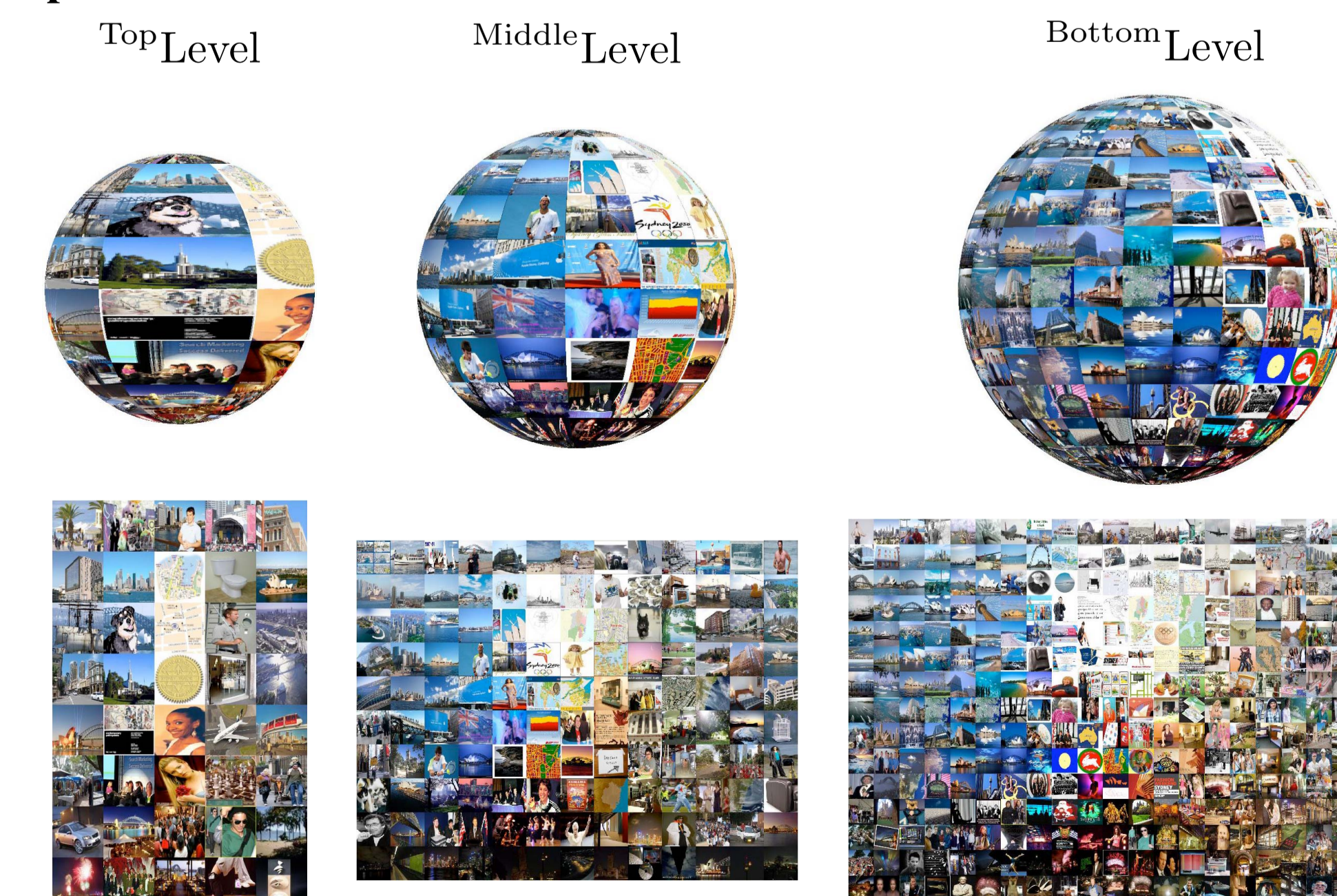


### Hierarchical Search System

Hierarchy of Grids

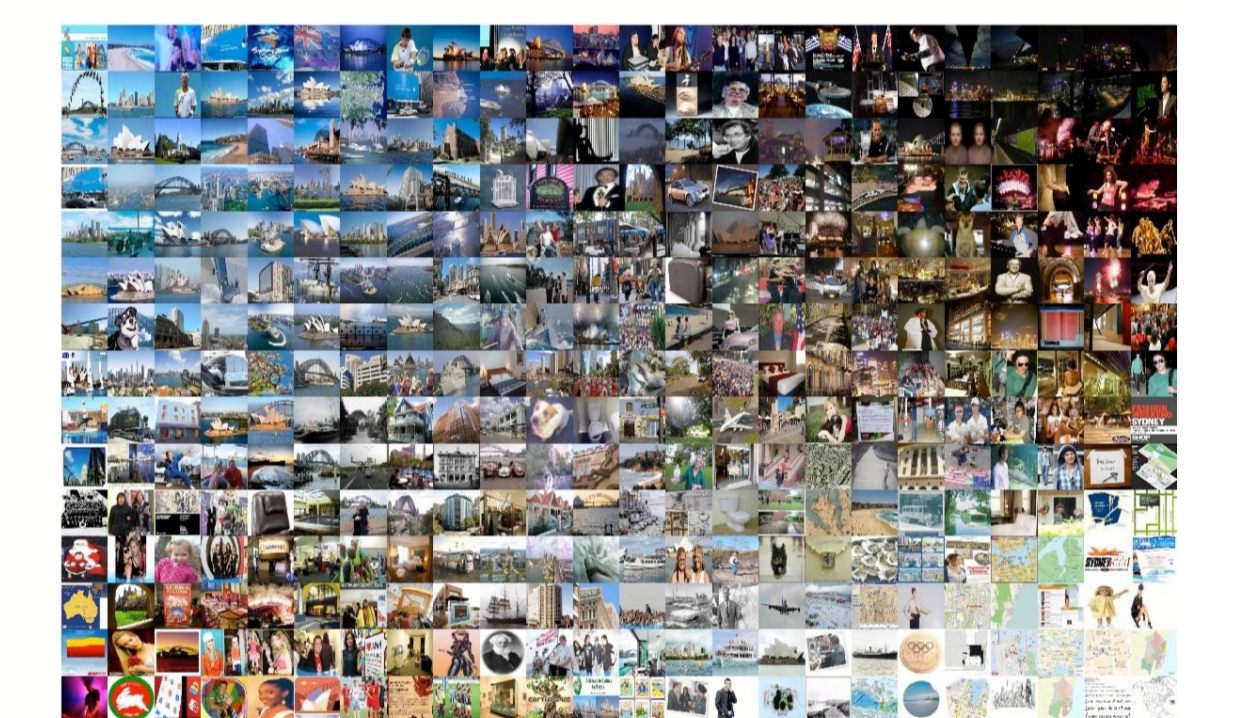
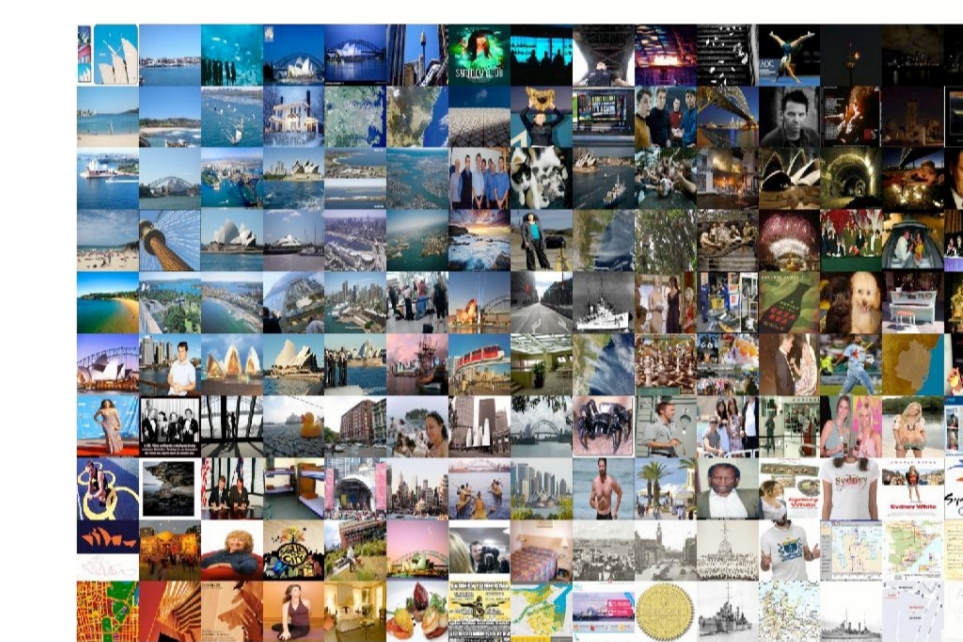
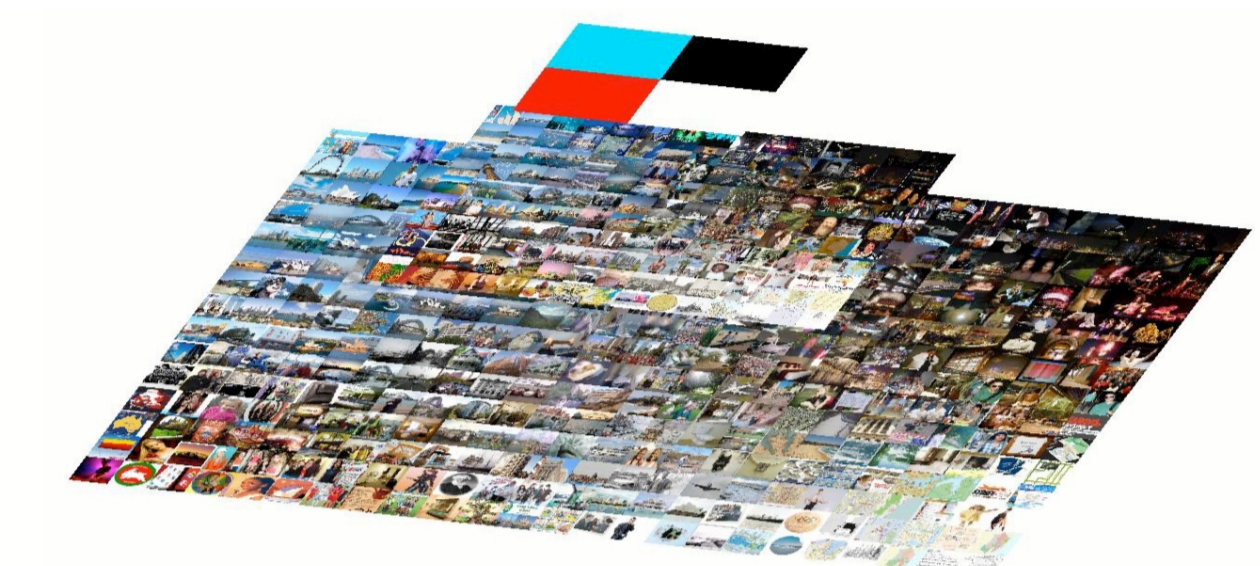


Hierarchy of Spheres



## Hierarchical Search System with Preferences

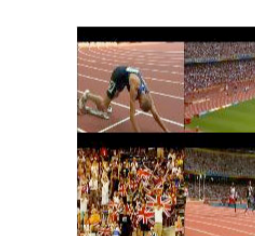
Color Preference (with Lab color space features)



Middle Level

Bottom Level

Semantic Preference (with SIFT features (Lowe 2004))



Queries



Middle Level

Bottom Level

## References

- N. Quadrianto, A. J. Smola, L. Song, and T. Tuytelaars. Kernelized sorting. *IEEE Trans. on Pattern Analysis and Machine Intelligence*, (in press), 2009.
- D. G. Lowe. Distinctive image features from scale-invariant keypoints. *IJCV*, 60:91–110, 2004.