

Rasterized Image Databases for Image Compression

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6.830
Database Systems
Final Project

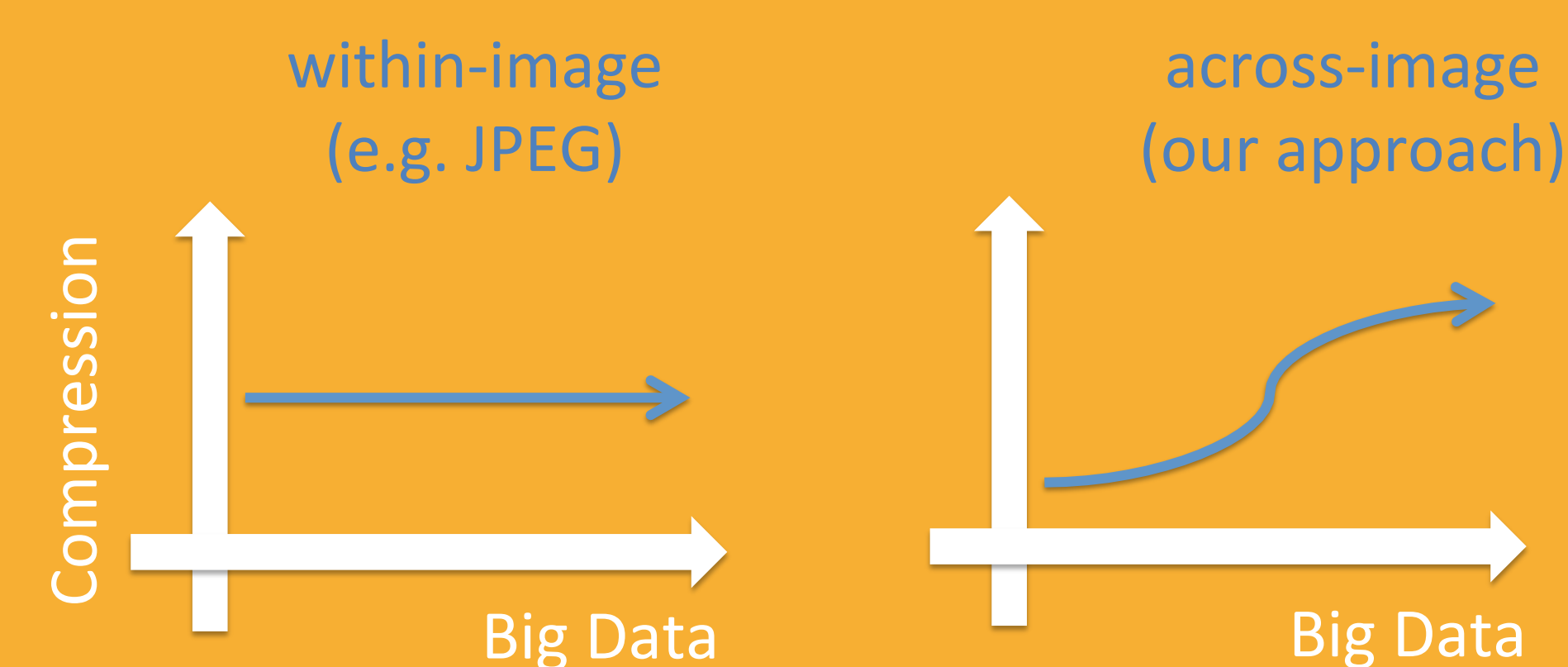
MOTIVATION

- More than 1.8 billion images uploaded to the internet every day
- Redundancy in large image collections can lead to more efficient storage

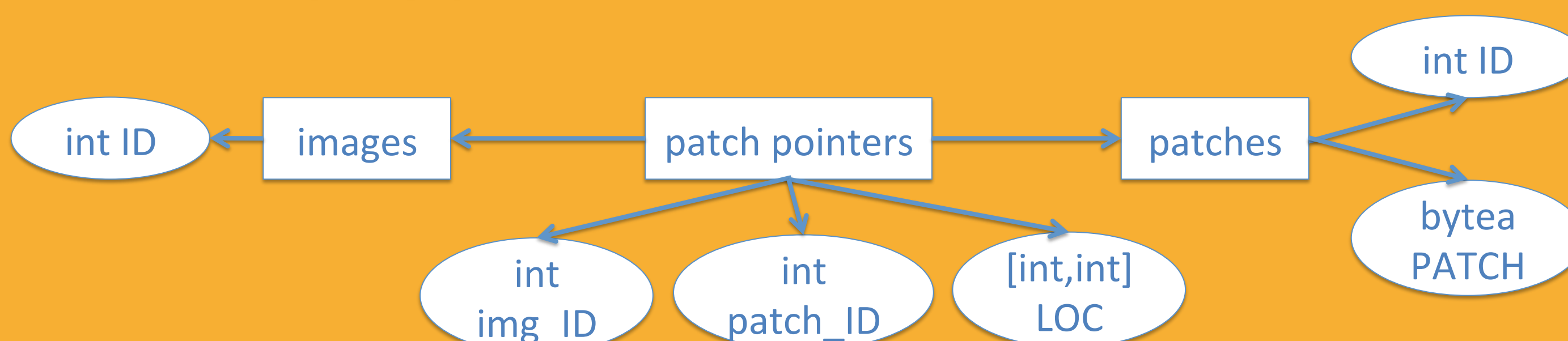


PATCH-BASED IMAGE CODING

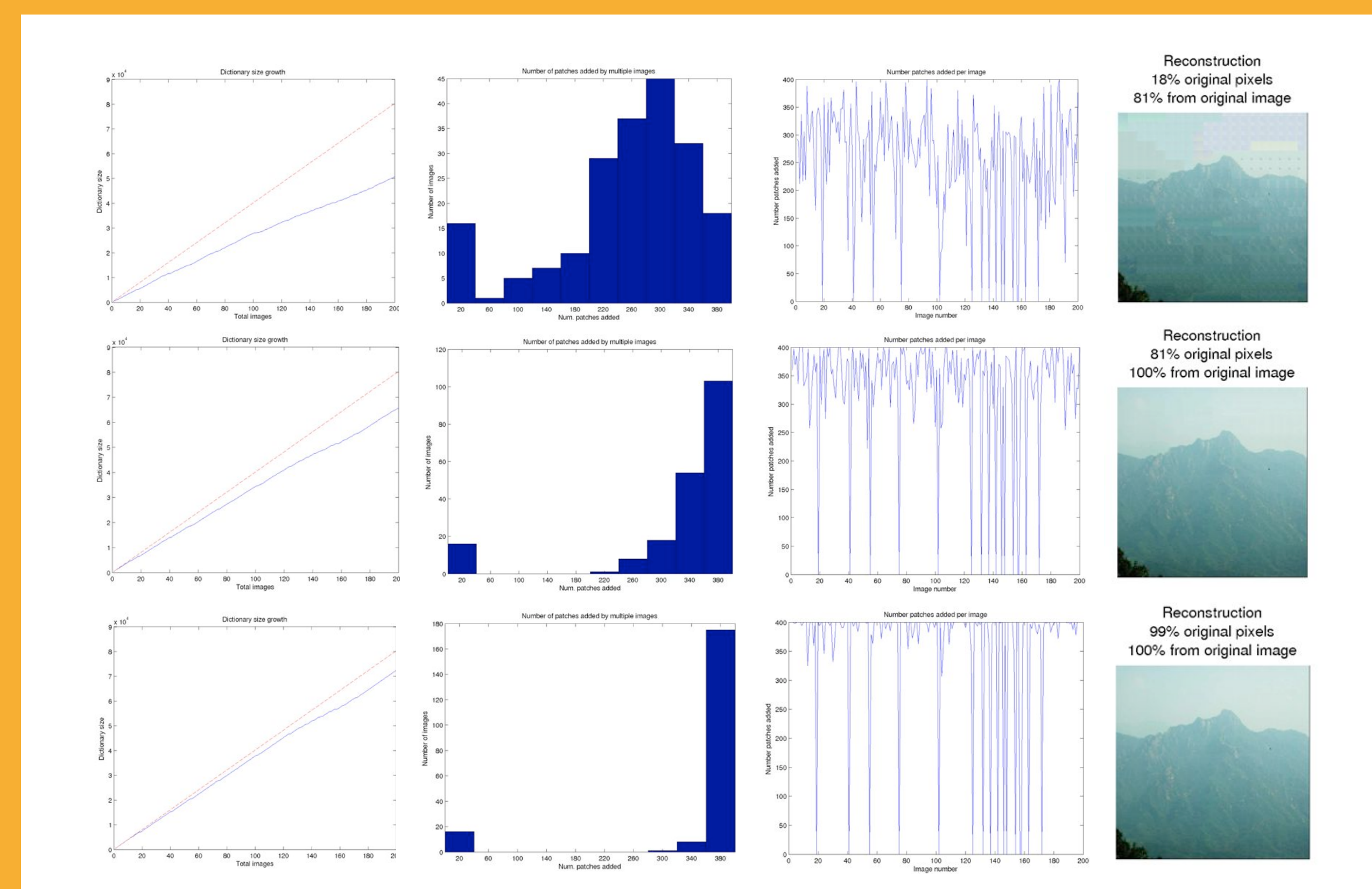
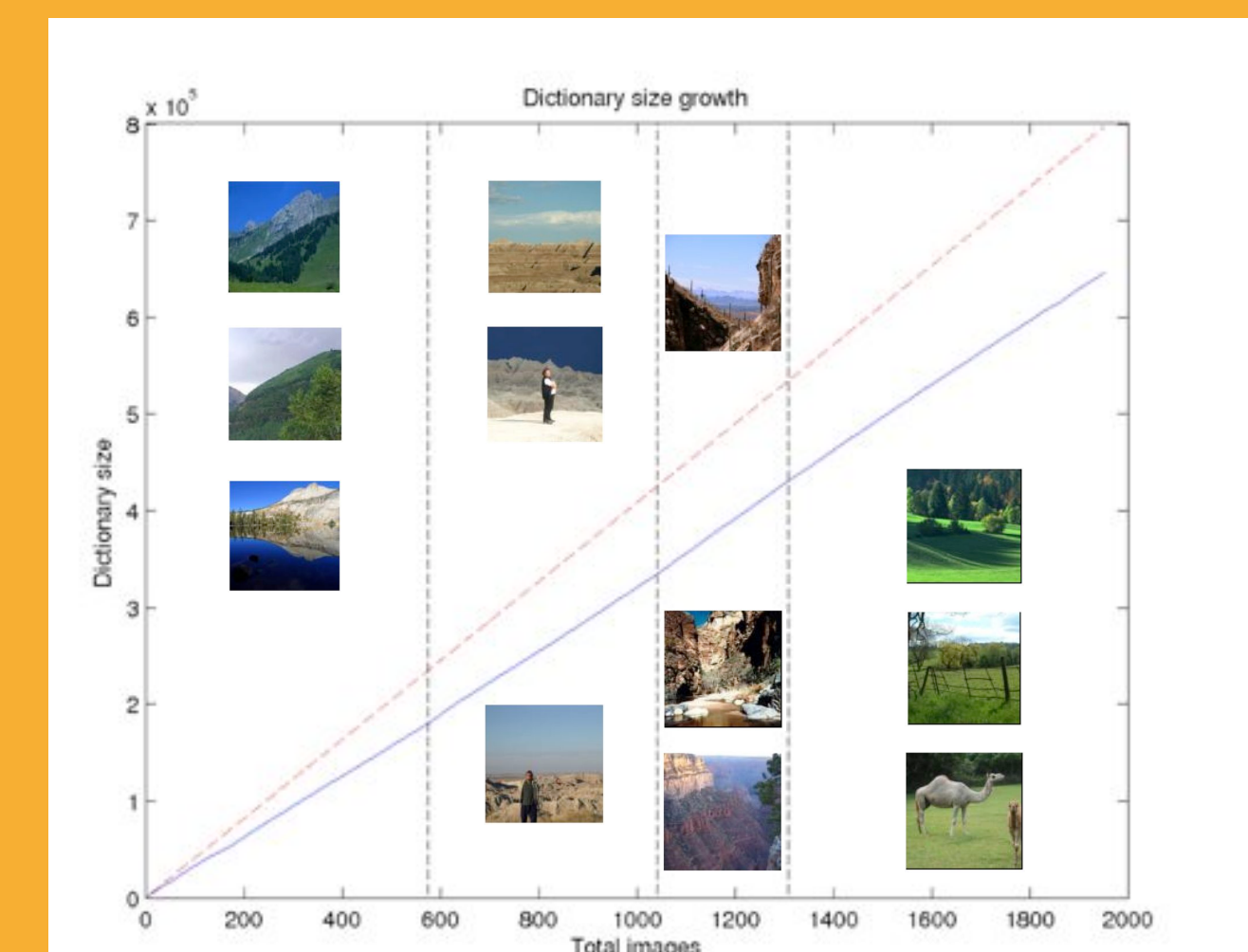
- Lossy compression
- Compression savings INCREASE as the database size INCREASES
- Rasterization is a simple operation – easily deconstruct and reconstruct images using patches



DATABASE SCHEMA



PARAMETER SELECTION



RESULTS

APPLICATIONS

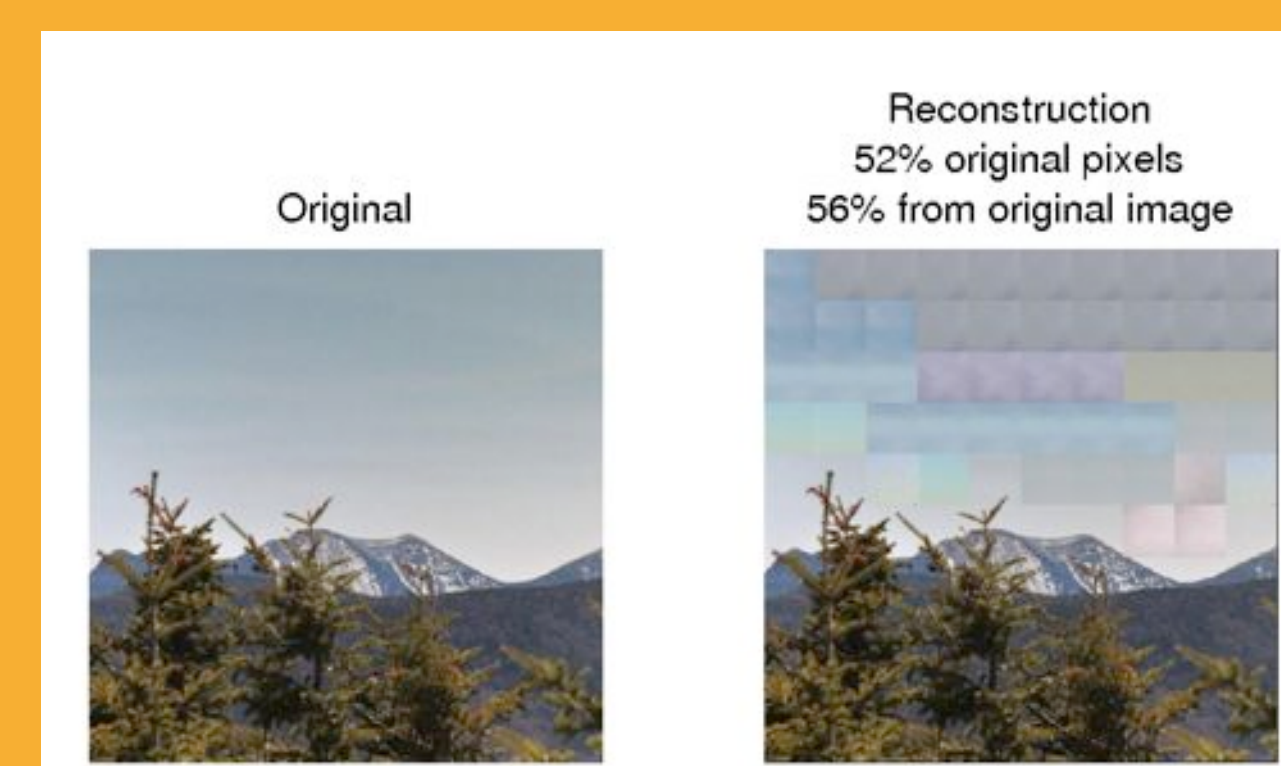
APPROACH

Algorithm 1 Insert Image I into database
1: $Patches \leftarrow \text{CutIntoPatches}(I, \text{patch_size}=n)$
2: **for** P in $Patches$ **do**
3: $SimPat \leftarrow \text{FindLikelySimilarPatches}(P)$
4: $P_{closest} \leftarrow \text{argmin}\{S(P, P_i)\}$
5: **if** $\text{then } S(P, P_i) > T$
6: insert P into patches

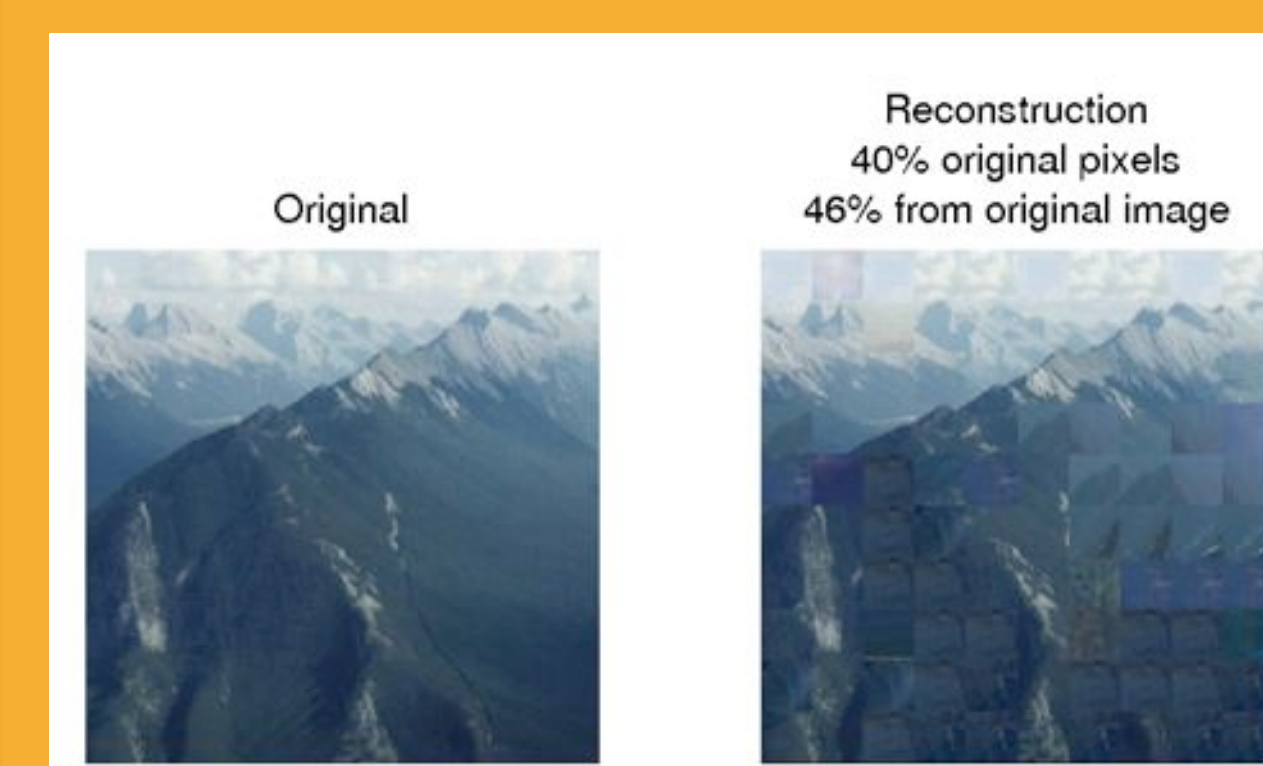
Computing similarity per color channel: $S(P_1, P_2, i) = \frac{\|P_1(i) - P_2(i)\|^2}{n^2}$

- CIE (LUV) color space
- Threshold per color channel: $S(P_1, P_2, i) < T_i$

The importance of the right color space and a thresholding each color channel:



The importance of the right similarity function and threshold:



The importance of the right patch size:

