

# Comprehensive Report: Jigsaw Puzzle Image Assembly Pipeline



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## I. Introduction

This report details the design, methodology, and rationale behind the **Jigsaw Puzzle Image Processing & Assembly Pipeline**, which encompasses **Milestone 1 (Image Processing & Feature Preparation)** and **Milestone 2 (Puzzle Solving & Reconstruction)**. The entire, authoritative implementation for both stages is consolidated within the `Milestone1notebook.ipynb`. The project's core objective is to achieve **deterministic, training-free** reconstruction of grid-based jigsaw puzzles (2x2, 4x4, and 8x8) by transforming a raw image into a standardized set of pieces, and then utilizing a global optimization algorithm to restore the original image structure.

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## II. Overall Flow and What Happens

The complete system follows a clear, sequential pipeline:

Step	Milestone	Action/Logic	Purpose (Contribution to Goal)
1-3	Milestone 1	<b>Structural Standardization &amp; Slicing:</b> Raw images are resized to a fixed 224x224 pixels and uniformly sliced into $N^2$ discrete tiles (4, 16, or 64 pieces).	Provides a <b>clean, standardized input</b> and the essential <b>ground truth</b> (known correct arrangement) for the solver.
4	Milestone 2	<b>Tile Representation:</b> Pieces are represented by their raw image patches, relying on <b>patch-level color information</b> .	Preserves all visual information necessary for reliable comparison while avoiding information loss.
5	Milestone 2	<b>Global Matching:</b> Puzzle pieces are compared using mean absolute difference in <b>LAB color space</b> . The resulting cost matrix is solved by the <b>Hungarian Algorithm</b> (linear sum assignment).	Produces a <b>globally optimal one-to-one assignment</b> , preventing local placement errors.
6	Milestone 2	<b>Reconstruction:</b> Matched tiles are placed exactly once into their assigned grid positions to form the final reconstructed image.	Ensures deterministic and repeatable restoration of the original image structure.

### III. Milestone 1: Slicer Logic and Method Rationale

The role of Milestone 1 was intentionally **restricted to deterministic image standardization and slicing only**. No perceptual filtering or feature extraction is performed at this stage, ensuring a clean, lossless representation.

#### Flow and Slicer Logic

The PuzzleSlicer component performs the following:

1. **Fixed Global Resize:** Input images are resized to **224x224 pixels** using area-based interpolation. This guarantees uniform spatial dimensions, eliminating scale-related ambiguity during reconstruction.
2. **Deterministic Grid Slicing:** Images are sliced into uniform  $N \times N$  grids using fixed index arithmetic. This produces perfectly aligned, non-overlapping tiles and preserves the true spatial relationships.

#### Comparison with Alternative Milestone 1 Approaches

The final Milestone 1 approach (deterministic resizing and slicing) was selected over earlier, more complex preprocessing techniques due to its superior reliability.

Final Method Advantage	Alternative Approaches (Rejected)	Rationale for Rejection
Lossless Data Preservation	Perceptual Preprocessing (e.g., Mean Shift Filtering, CLAHE, Canny Edge Detection)	Filtering operations caused <b>irreversible information loss</b> and introduced sensitivity to noise/illumination, negatively affecting downstream matching.
Structural Consistency	Contour-Based Representations	Contour-based features increased implementation complexity without providing consistent accuracy gains for rectangular, grid-based pieces.

The slicer logic, by creating standardized and perfectly aligned tiles, provides the essential **structurally consistent puzzle representation** that the Milestone 2 solver relies on.

## IV. Milestone 2: Algorithm and Method Rationale

Milestone 2 performs the final tile-to-slot assignment, using a combination of a perceptually accurate cost metric and a global optimization method.

### Key Methods

1. **Patch-Based LAB Color Matching:** The cost metric uses the mean absolute difference in the **LAB color space**. This space correlates better with human perceptual similarity, making the matching metric **robust against illumination changes**.
2. **Hungarian Algorithm (Linear Sum Assignment):** This globally optimal assignment algorithm is used to solve the cost matrix.

### Comparison with Alternative Milestone 2 Strategies

The combination of LAB matching and the Hungarian Algorithm was selected to optimize for **accuracy, robustness, interpretability, and computational efficiency**.

Final Method Advantage	Alternative Algorithms (Rejected)	Rationale for Selection
Global Optimality	Greedy and Local Matching Methods	Greedy approaches suffer from <b>cascading errors</b> and produce locally optimal but globally incorrect assemblies. The Hungarian Algorithm guarantees a globally optimal assignment, minimizing total error.
Color-Based Robustness	Edge-Based and Contour-Based Solvers	Edge-based methods are unreliable for grid-based puzzles due to limited geometric information and sensitivity to noise. <b>LAB color similarity</b> provides a stronger, more stable cue for rectangular pieces.
Deterministic & Simple	Feature-Descriptor Methods (SIFT, ORB) or Learning-Based Methods	Feature-descriptor methods are complex and require strong local features. Learning-based methods require large datasets and lack interpretability. The adopted solution is <b>training-free</b> , transparent, and reproducible.

## V. Conclusion

The Jigsaw Puzzle Assembly Pipeline successfully integrates a minimalist, structurally precise front-end (Milestone 1) with a globally optimal back-end (Milestone 2).

By restricting Milestone 1 to **lossless standardization and deterministic slicing**, the system ensures that Milestone 2 receives a perfectly prepared data set. The subsequent choice of **Patch-Based LAB Color Matching** combined with the **Hungarian Algorithm** addresses the inherent weaknesses of local heuristics, delivering a globally optimal reconstruction solution. This methodology achieves the project's goal of a **deterministic, interpretable, and highly accurate** puzzle solver across various grid complexities, reliably restoring the original image structure.

## VI. Github link

[loaymahmoud10/jigsaw-puzzle-processing: Milestone 1: High-Fidelity Image Processing Pipeline for Puzzle Tiles.](#)