PatchMatch: A Randomized Correspondence Algorithm for Structural Image Editing

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Structural Image Editing

Synthesize complex texture and image structures that resembles input imagery

Image retargeting

- Image completion
- Image reshuffling





(a) input

(b) hole and guides

(c) completion result



Matching Patches

• Nearest Neighbor Search





Which patch is most similar?

Matching Patches

• Naïve Approach



Sample every possible patch to find best match!

O(mM²)

Which patch is most similar?

• Search space

Patch offsets vs Patches

• Neighboring pixels have coherent matches

 Large number of random sampling will yield some good guesses.

Patches

VS.

Patch offsets





(x and y displacements)

Patch p offset search with 2 dimensions

Coherent matches with neighbors



Large numbers of guesses

M number of total pixels

Probability of correct random guess: 1/M

Probability of incorrect random guess: 1 - 1/M

Probability of all pixels with incorrect guess: $(1 - 1/M)^{M}$ [approximately 0.37]

 \Rightarrow Probability of at least 1 pixel with correct guess : 1 - $(1 - 1/M)^M$

 \Rightarrow Probability of at least 1 pixel with good enough guess: 1 - (1 - C/M)^M

Algorithm – 3 steps





Algorithm – Initialization

• Each pixel is given a random patch offset as initialization



Algorithm – Propagation

 Each pixels checks if the offsets from neighboring patches give a better matching patch. If so, adopt neighbor's patch offset.



Algorithm – Search

- Each pixels searches for better patch offsets within a concentric radius around the current offset.
- The search radius starts with the size of the image and is halved each time until it is 1.



Algorithm

- 1. Initialize pixels with random patch offsets
- 2. Check if neighbors have better patch offsets
- Search in concentric radius around the current offset for better better patch offsets
- 4. Go to Step 2 until converge.

O(mMlogM)



Algorithm



Speed Improvements

	Time [s]		Memory [MB]	
Megapixels	Ours	kd-tree	Ours	kd-tree
0.1	0.68	15.2	1.7	33.9
0.2	1.54	37.2	3.4	68.9
0.35	2.65	87.7	5.6	118.3

Impact

- Not only used in graphics, but in vision
 - Non-local means denoising
 - Image forensics
 - Object detection
- Video Tapestries





- Videos: Patchmatch in 3D
 - Temporal super-resolution



Results



(a) input

(b) result

Results



(a) input

(b) hole and guides

(c) completion result

Results



Results (Failure)





Results (Failure)



