

Report

Part 1:

Experiment 1:

Parameters:

PatchMatch parameters:

Initial NNF:	Generated internally
Iterations:	4
Patch size:	7
NLM h:	25.0
Alpha:	0.5
W:	300
K:	5
Run NLM algorithm:	True
Propagation enabled:	True
Random search enabled:	True

Output path and base filename: ../results/report/report

Visualization parameters:

Output files:	correspondences, color nnf, rec'd source
NNF subsampling:	100
NNF line width:	0.5
NNF line color:	r
Matplotlib server mode:	False
Tmp directory:	./

Last Iteration:

NNF :



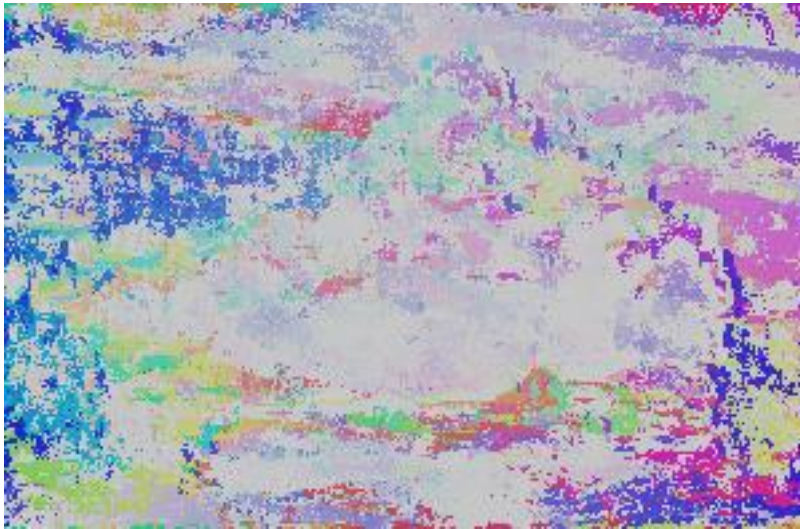
order=0

NNF:



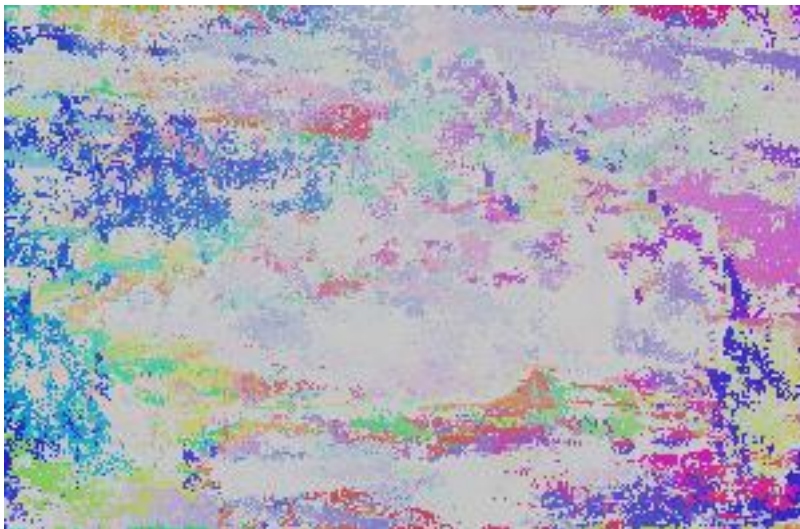
order=1

NNF:



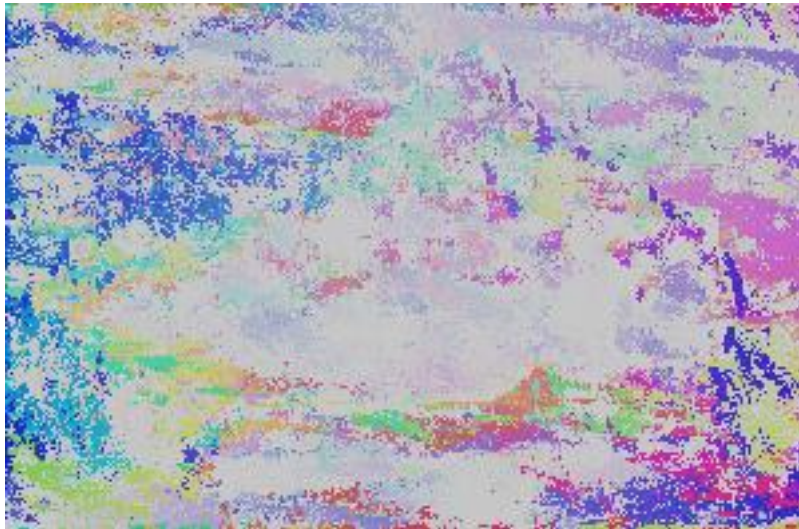
order=3

NNF:



order=4

NNF:



Experiment 2:

Parameters:

PatchMatch parameters:

Initial NNF:	Generated internally
Iterations:	4
Patch size:	7
NLM h:	25.0
Alpha:	0.5
W:	300
K:	3
Run NLM algorithm:	True
Propagation enabled:	True
Random search enabled:	True

Output path and base filename: ../results/report1/report1

Visualization parameters:

Output files:	correspondences, color nnf, rec'd source
NNF subsampling:	100
NNF line width:	0.5
NNF line color:	r
Matplotlib server mode:	False
Tmp directory:	./

Iteration4:

NLM :



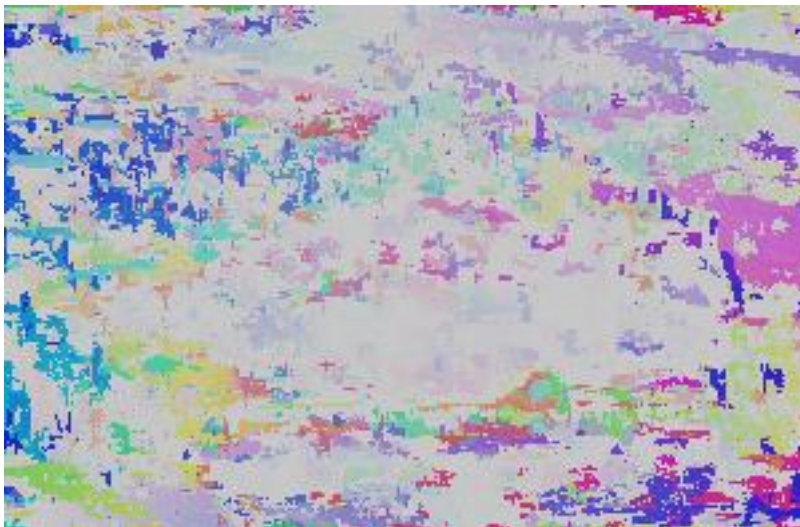
order=0

NNF:



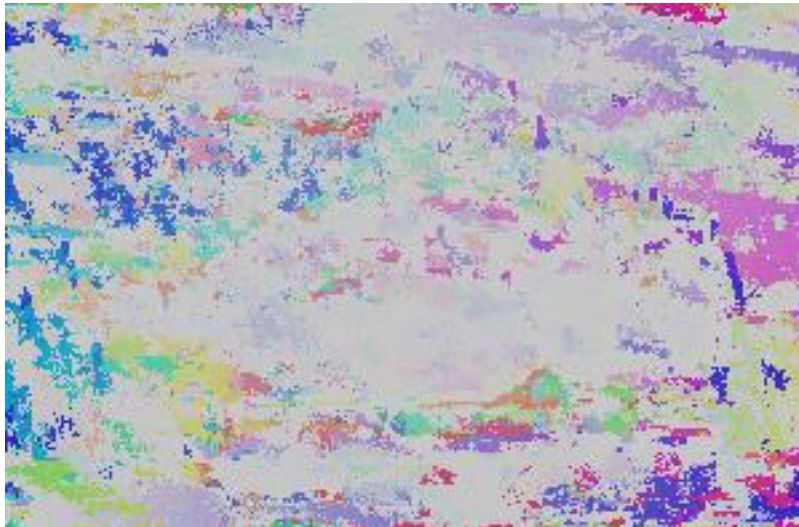
order=1

NNF:



order=2

NNF:



Compare Experiment 1 and Experiment 2. We can see that NLM in 1 is better than 2. Since order k in Experiment 1 is larger than Experiment 2, the proportion for K best NNF images are more balance. When the algorithm reconstructs a NLM image, for a noise pixel, we now have k similar pixels found on the image to reconstruct. For those samples, we use weight to balance each other. Thus, a noise pixel will be reduced by consider more k NNF.

Experiment 3:

Parameters:

```
PatchMatch parameters:
  Initial NNF:          Generated internally
  Iterations:           4
  Patch size:           19
  NLM h:                25.0
  Alpha:                0.5
  W:                    300
  K:                    5
  Run NLM algorithm:    True
  Propagation enabled:  True
  Random search enabled: True
Output path and base filename: ../results/report3/report3
Visualization parameters:
  Output files:          correspondences, color nnf, rec'd source
  NNF subsampling:       100
  NNF line width:        0.5
  NNF line color:        r
  Matplotlib server mode: False
  Tmp directory:         ./
```

Last Iteration

NLM :



order=0

NNF:



order=1

NNF:



order=2

NNF:



order=3

NNF:



order=4

NNF:



In Experiment 1 and 3, for each iteration, NNFs is unlikely to change when patch size is bigger. Because patch size determines the score of similarity, and if patch size is big, at

propagation, for each pixel, the nearby patches have the same rating, so offset is unlikely to change at every iteration. As a result, the NLM image twists.

Experiment 4:

Parameters:

```
PatchMatch parameters:
  Initial NNF:           Generated internally
  Iterations:            4
  Patch size:            7
  NLM h:                 25.0
  Alpha:                 0.5
  W:                    15
  K:                     5
  Run NLM algorithm:     True
  Propagation enabled:   True
  Random search enabled: True
Output path and base filename: ../results/report2/report2
Visualization parameters:
  Output files:           correspondences, color nnf, rec'd source
  NNF subsampling:        100
  NNF line width:         0.5
  NNF line color:         r
  Matplotlib server mode: False
  Tmp directory:         ./
```

Iteration 4:

NLM :



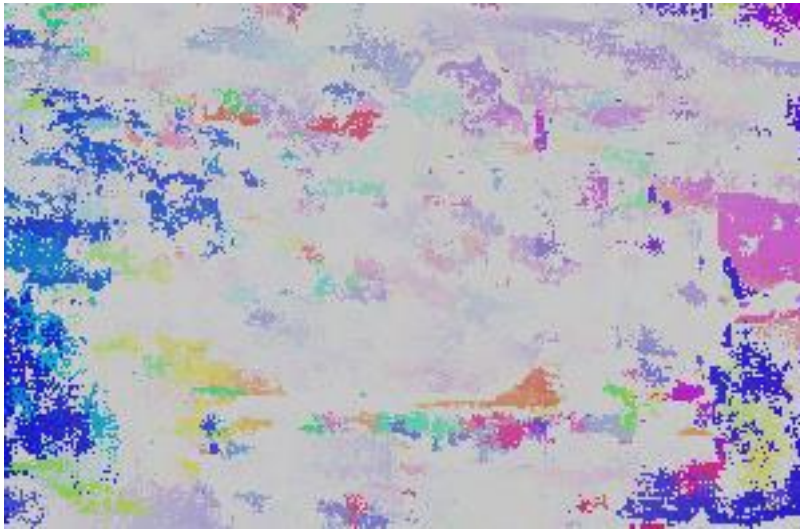
order=0

NNF:



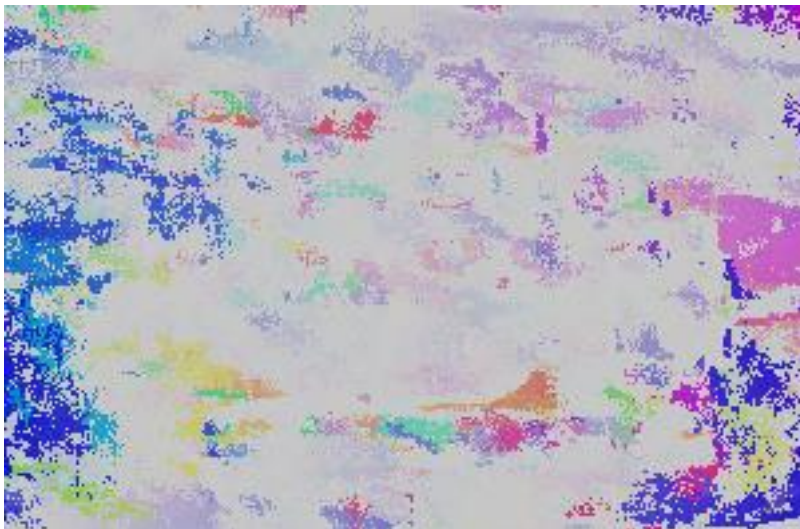
order=2

NNF:



order=3

NNF:



order=4

NNF:



In this Experiment, the random search patch radius is significant smaller than the Experiment 1. When the radius is small, NNF pictures tend to grey, because the random search size is small and the algorithm has fewer sample to check. Thus, many pixels may trap in a local min area, which leads NNF pictures to grey.

Generalized PatchMatch on the jaguar2 image pair for $k = 3$:
(source.png and target.png)

Iteration 1:

order=0

NNF:



NNF-vector:



Rec-source:



order=1

NNF:



NNF-vector:



Rec-source:



order=2

NNF:



NNF-vector:



Rec-source:



Iteration 2:

order=0

NNF:



NNF-vector:



Rec-source:



order=1

NNF:



NNF-vector:



Rec-source:



order=2

NNF:



NNF-vector:



Rec-source:



Iteration 3:

order=0

NNF:



NNF-vector:



Rec-source:

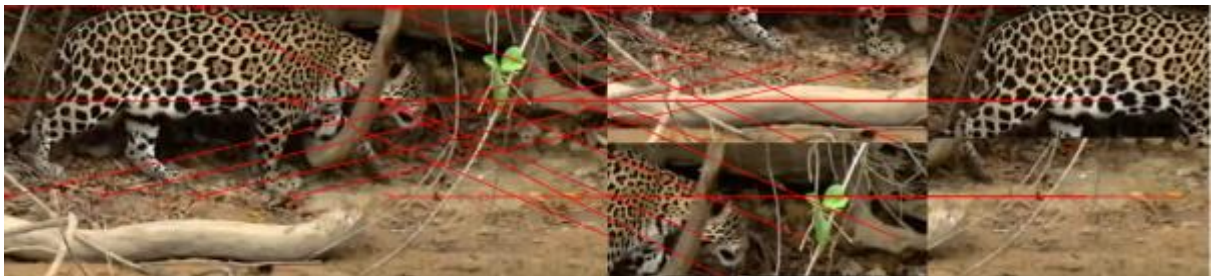


order=1

NNF:



NNF-vector:



Rec-source:



order=2

NNF:



NNF-vector:



Rec-source:



Iteration 4:
order=0
NNF:



NNF-vector:



Rec-source:



order=1

NNF:



NNF-vector:



Rec-source:

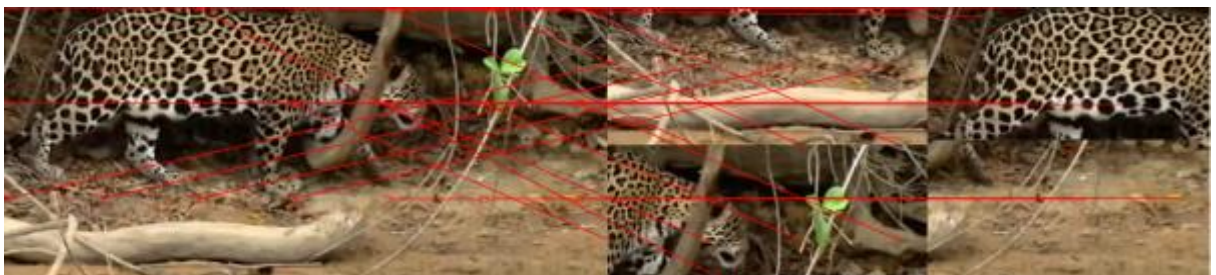


order=2

NNF:



NNF-vector:



Rec-source:

