# Efficient palette-based decomposition and recoloring of images via RGBXY-space geometry 

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## Motivation: Layers Organize Images

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Goal

Goal

## Layers



Goal
Reconstruction

## Layers


$\square$


## Subproblems

1. Palette extraction

2. Palette-based layer decomposition


## Related Work

- Palette extraction for image editing
- Shapira et al. [2009]
- O'Donovan et al. [2011]
- Lin et al. [2013]
- Gerstner et al. [2013]
- Chang et al. [2015]
- Tan et al. [2016]



## Related Work

- Order-dependent translucent layers
- Richardt et al. [2014]
- Tan et al. [2015]
- Tan et al. [2016]


Decomposing Images into Layers via RGB-space Geometry [Tan et al. 2016]

## Related Work

- Order-independent additive-mixing layers
- Lin et al. [2017]; Zhang et al. [2017], Aksoy et al. [2017].


Unmixing-Based Soft Color Segmentation for Image Manipulation [Aksoy et al. 2017]

## Related Work

- Physically-based layers
- Abed et al. [2014]; Tan et al. [2015]; Aharoni-Mack et al. [2017]; Tan et al. [2018].


Pigmento: Pigment-Based Image Analysis and Editing [Tan et al. 2018]

## Our approach



## Our approach

- Geometry-based convex palettes



## Our approach

- Geometry-based convex palettes
- Simpler



## Our approach

- Geometry-based convex palettes
- Simpler
- More general



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- Geometry-based convex palettes
- Simpler
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- Additive-mixing layers



## Our approach

- Geometry-based convex palettes
- Simpler
- More general

- Additive-mixing layers
- Single colors



## Our approach

- Geometry-based convex palettes
- Simpler
- More general

- Additive-mixing layers
- Single colors
- More general



## Palette extraction

## Convex hulls in RGB

- Image colors show a convex structure in RGB [Tan et al. 2016]



## Palette Size

- The convex hull can be simplified to any complexity level.



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## Palette Size

- Our automatic error-bound simplification



## Image Decomposition

## Extracting mixing weights

image


RGB-space


Optimization

## Extracting mixing weights

image


RGB-space


- Slow for high resolutions

Optimization

## Extracting mixing weights

image


RGB-space


- Slow for high resolutions
- Many parameters to tune


## Optimization

## Extracting mixing weights

image


RGB-space


- Slow for high resolutions
- Many parameters to tune
- Per-image parameters


## Optimization

## Extracting mixing weights

image

palette

RGB-space


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## Extracting mixing weights

image

palette

RGB-space


Optimization

- Slow for high resolutions
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- Per-image parameters



## Extracting mixing weights

image


RGB-space


Optimization
palette

## Extracting mixing weights



## Extracting mixing weights



## Extracting mixing weights



## Extracting mixing weights



RGB-space


- Fast
- No parameters to tune
- Does not guarantee spatial smoothness


## Extracting mixing weights

image

palette

RGB-space


- Fast
- No parameters to tune
- Does not guarantee spatial smoothness



## Extracting mixing weights

image

palette

RGB-space


Optmizaton
Generalized Barycentric Coordinates

- Fast
- No parameters to tune
- Does not guarantee spatial smoothness



## Extracting mixing weights



## Extracting mixing weights











## Two-level decomposition

image

RGB palette

## Two-level decomposition



## Two-level decomposition



## Two-level decomposition



## Delaunay Tessellation in RGBXY space

Extract barycentric mixing weights $\mathbf{W}_{\text {RGBXY }}$

## Delaunay Tessellation in RGBXY space



Extract barycentric mixing weights $\mathbf{W}_{\text {RGBXY }}$

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Extract barycentric mixing weights $\mathbf{W}_{\text {RGBXY }}$

## Delaunay Tessellation in RGBXY space



Extract barycentric mixing weights $\mathbf{W}_{\text {RGBXY }}$

## Two-level decomposition

RGB palette
RGBXY vertices
(projected to RGB)

image

## Two-level decomposition



## Tessellation in RGB space



Extract barycentric mixing weights $\mathbf{W}_{\mathbf{R G B}}$

## Tessellation in RGB space



Extract barycentric mixing weights $\mathbf{W}_{\mathbf{R G B}}$

## Tessellation in RGB space



Extract barycentric mixing weights $\mathbf{W}_{\mathbf{R G B}}$

## Tessellation in RGB space



Extract barycentric mixing weights $\mathbf{W}_{\mathbf{R G B}}$

## Tessellation in RGB space



Extract barycentric mixing weights $\mathbf{W}_{\mathbf{R G B}}$

## Two-level decomposition



## Two-level decomposition

image

## Tessellation in RGB space



## Tessellation in RGB space



## Tessellation in RGB space






## Two-level decomposition

image

RGB palette

## $\mathrm{W}=\mathrm{W}_{\mathrm{RGB}}{ }^{*} \mathrm{~W}_{\mathrm{RGBXY}}$

## Two-level decomposition

image

RGB palette
Palette updates
Fixed
$\mathrm{W}=\mathrm{W}_{\mathrm{RGB}}{ }^{*} \mathrm{~W}_{\mathrm{RGBXY}}$

## Two-level decomposition

image


## Two-level decomposition

image


## Two-level decomposition

image

RGB palette
Palette updates
Fixed
$\mathrm{W}=\mathrm{W}_{\mathrm{RGB}}{ }^{*} \mathrm{~W}_{\mathrm{RGBXY}}$


Updating $W_{R G B}$ is independent of image size.

## Two-level decomposition



Updating $\mathrm{W}_{\mathrm{RGB}}$ is independent of image size.
Other methods need to re-compute everything from scratch.

## Performance



## Performance



## Performance



## Python Implementation

```
from numpy import * 
from scipy.spatia import ConvexHal
def RGBXY_weights( RGB_palette, RGBXY_data )
    RGBXY_hull_vertices = RGBXY_data[ ConvexHull( RGBXY_data ).vertices
    N_RGBXY = Delaunay_coordinates( RGBXY_hull_vertices, RGBXY_data )
    # Optional: Project outside RGBXY_hull_vertices[:,:3] onto RGB_palette convex hul..
    W_RGB = Star_coordinates( RGB_palette, RGBXY_hull_vertices[:,:3]
    def Star_coordinates( vertices, data )
    ## Find the star vertex
    star = argmin( linalg.norm( vertices, axis=1 ))
    ## Make a mesh for the palette
    hutl = ConvexHull( vertices
    simplices = [ [star] + list(fa the convex hul
    cluplos}=[[star] + list(face) for face in hull.simplices if star not in face ]
    barycoords = -1*ones( (data.shape[0], len(vertices))
    for s in simplices:
        s0 = vertices[s[:1]]
        b=linalg.solve( (vertices[s[1:]]-s0).T, (data-s0).T )
        b= append( 1-b.sum(axis=1)[:,None], b, axis=1 )
        ask bate bary aorlas when
        ask = (b>=0).all (axi
        barycoords[ix_(mask,s)] = b[mask]
        eturn barycoords
def Delaunay_coordinates( vertices, data ): # Adapted from Gareth Rees
    tri = Delaunay(vertices)
# Find the tetrahedron containing each target (or -1 if not found)
simplices = tri.find_simplex(data, tol=1e-6)
assert (simptices != -1).al() # data contains outside vertices.
# Affine transformation for simplex containing each datum
X = tri.transform[simplices,:data.shape[1]]
Y = data - tri.transform[simplices, data.shape[1]]
# Compute the barycentric coordinates of each datum in its simplex
barycoords = c_[b,1-b.sum(axis=1)]
barycoords = c_[b,1-b.sum(axis=1)]
rows = repeat(arange(len(data)).reshape((-1,1)), len(tri.simplices[0]), 1).ravel()
cols = tri.simplices[simplices].ravel()
vals = barycoords.ravel()
vals = barycoords.ravel()
```


## Comparisons

## Layer quality comparison with [Aksoy et al. 2017]



## Layer quality comparison with [Aksoy et al. 2017]



Recoloring comparison with three previous methods


Original


Ours

Recoloring comparison with three previous methods


Original
Aksoy et al. 2017


Ours

Recoloring comparison with three previous methods


Original
Aksoy et al. 2017


Ours

Recoloring comparison with three previous methods


Ours

Recoloring comparison with three previous methods


Ours

Recoloring comparison with three previous methods


Recoloring comparison with three previous methods


## Demo

## Javascript + Python with PyOpenCL

## Layer creation from scratch

```( colorful
```


## dd Random Palette Color



```( colorful
```


## dd Random Palette Color



## Layer creation from an automatic palette



Choose File No file chosen

- Prescribed number of layers:


## 6

 Add Random Palette Color


Choose File No file chosen

- Prescribed number of layers:


## 6

 Add Random Palette Color

## Interactive decomposition gives more control to the users



## Interactive decomposition gives more control to the users



## Interactive decomposition gives more control to the users



## Conclusion

- An extremely efficient approach to layer decomposition via RGBXY geometry



## Conclusion

- Our two-level decomposition supports real-time decomposition when palette editing.


## Palette updates

Fixed

## W = <br> Whgb <br> * $\mathrm{W}_{\mathrm{Rg}} \mathrm{xy}$

## Conclusion

- It's important to capture the "line of greys".



## Limitations

- In isolated cases, the 5D convex hull takes somewhat longer than usual to compute.



## Limitations

- Our star tessellation assumes that palette colors are vertices of a convex polyhedron.
- For palette colors in the interior, must use inferior Delaunay tessellation.


Star teosellation


Delaunay tessellation

## Future Work

- More speed via super-pixels or parallel convex hull algorithms.



## Future Work

- Robustness via approximate convex hull algorithms.



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- Robustness via approximate convex hull algorithms.



## Future Work

- Robustness via approximate convex hull algorithms.



## Thank You!

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- Jianchao Tan: jtan8@gmu.edu
- Jose Echevarria: echevarr@adobe.com
- Yotam Gingold: ygingold@gmu.edu
- Project Website (GUI, code, data): https://cragl.cs.gmu.edu/fastlayers/
- Artists: Adelle Chudleigh; Dani Jones; Karl Northfell; Michelle Lee; Adam Saltsman; Yotam Gingold;

DeviantArt user Sylar113; Fabio Bozzone; Piper Thibodeau; Spencer Nugent; George Dolgikh; DeviantArt user Ranivius.

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- NSF, Adobe, Google.


## Extra slides

## Possible questions

- Star triangulation starting from black palette color, what if no black color in extracted palette?
- Does your method require palette to cover all pixel colors when editing palette in GUI? What if I want some palette colors that is inside color point cloud?
- In your performance figure, there are one or two cases that are slower than many others. Can you describe the worst case performance of your method?
- Do you have failure case?
- How do you measure the quality of your layer results and your interactive editing GUI?

