Perspective Projection



Perspective Projection



Preview: Our Result



Preview: Our Result



Motivation

 Perspective projection makes unnatural look on faces near image corners.

- **Portrait** is important to photography.
 - 44% of photos contains faces.
 - E.g., selfie, group portraits, group, etc.



Group Selfie Cam

Google Pixel 3

Trend: Wide-angle Camera are GROWING!

Make	Samsung	Google	Asus	Huawei	LG	OnePlus	Sony	Xiaomi
Flagship Model (2019)	S10 plus	Pixel 3	Zenfone 6	P30 Pro	V50 Thin Q	7 Pro	Xperia 1	Mi9
Field-of-view (FOV)	123°	97°	125°	107°	107°	117°	130°	104°
35mm equiv. focal length	12mm	19mm	11mm	16mm	16mm	13mm	10mm	17mm

Human vision (FOV/35mm-equiv): 60°/37.5mm

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• Human vision (FOV/35mm-equiv): 60°/37.5mm

• Major camera phone prior to 2018 (FOV/35mm-equiv): 80°/26mm

NOT A Lens Distortion



Lens distorted image

NOT A Lens Distortion



Lens distorted image

Corrected image (perspective projection)

Goal: Distortion-Free Wide-angle Portraits

- Makes faces look natural even large at field-of-view (FOV).
 - As if faces are captured from the camera center.
 - Robust to face poses, facial expressions, and occlusions.

Goal: Distortion-Free Wide-angle Portraits

- Makes faces look natural at large field-of-view (FOV).
 - As if faces are captured from the camera center.
 - Robust to face poses, facial expressions, and occlusions.
- Works for mobile platform.
 - Automatic correction without user intervention.
 - Efficient processing to achieve interactive rate.

Related Work: Global Projection





Stereographic Projection



Mercator Projection

Related Work: Content-aware Warping

• Spatially-variant radial correction [Zorin and Barr 1995].



Input

Output

Related Work: Content-aware Warping

• Optimizing global conformality [Carroll et al. 2009].



Input [Carroll et al. 2009]

Related Work: Content-aware Warping

- Optimizing global conformality [Carroll et al. 2009].
- Selects a different virtual camera viewpoint [Tehrani et al. 2016].







Input

[Tehrani et al. 2016]

Input [Carroll et al. 2009]

Related Work: Perspective-aware Warping

• Corrects foreshortening [Fried et al. 2016]



Input

Output

Our Idea: Local Stereographic Projection on Faces







Perspective projection - Face is stretched 😕

- Background looks good 😊

Stereographic projection
- Face looks natural 😳

- Background is distorted 😕

Our result by local warp - Face look natural 😳

- Background look good 🙂



Input (perspective projection)



Subject segmentation using CNN [Wadhwa et al. 2018]



Face detector for face region



Input (perspective projection)

Stereographic projection

Mesh optimization

Output by warping

Naïve Mesh Blending



(a) Input (97° FOV)

Inset of (a)

Naïve mesh blending

Naïve Mesh Blending



(a) Input (97° FOV)

Inset of (a)

Naïve mesh blending

Optimized mesh warp

Mesh Optimization

• Defines an energy function over the vertices on the mesh.

$\{\mathbf{v}_i^*\} = \operatorname{argmin} E_t(\{\mathbf{v}_i\}), \\ \{\mathbf{v}_i\}$

Face Objective Term

• Each detected face k associates with a face term $E_{s,k}$,

$$E_{f} = \sum_{k} E_{s,k},$$

stereographic projection vertice
$$E_{s,k} = \sum_{i \in \mathbf{B}_{k}} w_{i} n_{i} \|\mathbf{v}_{i} - (\mathbf{S}_{i} \mathbf{u}_{i} + \mathbf{t}_{k})\|_{2}^{2} + \lambda(\mathbf{S}_{k}),$$

face weight; 1/0 for face/nonface region using segmentation.



Face Objective Term

$$(1 + \exp(-(r - r_a)/r_b))^{-1} \qquad \qquad \lambda(\mathbf{S}_k) = w_s ||a_k - s_t||_2^2,$$

9

$$E_{s,k} = \sum_{i \in \mathbf{B}_k} v_i m_i |\mathbf{v}_i - (\mathbf{S}_k \mathbf{u}_i + \mathbf{t}_k)||_2^2 + \lambda(\mathbf{S}_k)$$
$$\mathbf{S}_k = \begin{bmatrix} a_k & b_k \\ -b_k & a_k \end{bmatrix}$$

Benefits of Face Term





Input Without *S, t* Our method

Benefits of Subject Mask



(a) Input and mask

Inset of (a)

Without mask

Our method

Line Bending Terms

$$E_b = \sum_i \sum_{j \in N(i)} \|(\mathbf{v}_i - \mathbf{v}_j) \times \mathbf{e}_{ij}\|_2^2$$

$$E_r = \sum_i \sum_{j \in N(i)} \|\mathbf{v}_i - \mathbf{v}_j\|_2^2.$$



Inset of (a)

(a) Input

Without line bending term

Our method

Mesh Boundary Padding

• Relax boundary by padding additional vertices.



(a) Input Inset of (a) Without padding Our method

Mesh Optimization

• Total energy cost by summing up

- Face term
- Line bending term
- Regularization term
- Boundary conditions
- Solves the linear energy function with Ceres.

Handling Lens Distortion

• Combines the optimal mesh with lens undistortion warp.



(a) Lens undistorted input

Our method by warping (a)

Results

- 167 photos from Flickr.
- FOVs range from 70° -120°.

• Various features:

- Different genders, ages, facial expressions, lighting conditions.
- Glasses, hair styles, accessories.
- Available on our project website.

Input 97°FOV Pixel 3







Input <u>104°FOV</u> DSLR M

30



Input <u>97°FOV</u> Pixel 3







Input 97°FOV (Pixel 3)

No of Fi



Correction at Various FOVs

• Subjects at a 5x3 grid over a 97°FOV camera.



Input

Output

Normal FOV



Normal FOV



Output

Large Group



Large Group



Output

Input





Output





Stress Test



Input <u>103°FOV</u> (GoPro Hereo5)

Stress Test



Output

Occlusion



Occlusion



Output

Comparisons



Input (100° FOV) Pannini [Sharpless et al. 2010] Adobe Photoshop Perspective Warp Our method

Comparisons



Input (92° FOV) [Zorin and Barr 1995] Our method

Comparisons: DxO Viewpoint3



Input (97° FOV)

DxO Viewpoint3 Volume Deformation

Our method

Comparisons: Samsung S9+



Input (80° FOV)

Samsung S9+ "Face Shape Correction"

Our method

User Study on Amazon Mechanic Turk

- 1047 photos from Flickr and collected by us.
- FOVs range from 70°-120°.
- 5 testers on each photo.
 - "looks more natural with less distortion."



Comparator	Perspective projection	Stereographic projection	Mercator projection	Pannini projection
Favoring our method	92.4%	84%	81%	85%

Runtime

- Takes 920ms for a 12MP image on Qualcomm SDM845
 - Segmentation: 280ms (GPU)
 - Optimization: 340ms (CPU)
 - Warping: 115ms (GPU)

Limitation: missing face detection.



Input (104° FOV)

Output

Limitation: uncorrected body.

Input (90° FOV)



Output

Summary

- Introduce a distortion correction algorithm for wide-angle portraits.
- Based on mesh optimization and subject segmentation.
- Fully automatic and suited for mobile platform.
 - Basis for Wide-angle Group Selfie on Pixel 3.



Our method

Acknowledgement

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Thanks to Denis Zorin and all the photography models in this work for photo usage permission.



• Find us in Poster session for Live Demo on Pixel 3!

Wednesday, 31 July, 12:15-1:15 pm



Input (97° FOV)

Our method

Poster session for Live Demo on Pixel 3: Wed, 31 July, 12:15-1:15 pm



Input (perspective projection) 103° FOV Our corrected result