

Joint Super-Resolution and Optical Flow Estimation

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Outline

- 1 Introduction
- 2 Energy Minimization Approach
- 3 Results
- 4 Future Work



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- 1 Introduction
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Introduction

- Super-Resolution
 - Enhance resolution of images
 - Gain Additional Details



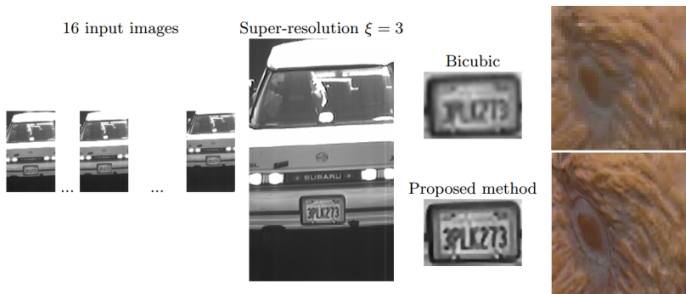
Introduction

- Super-Resolution
 - Naive Upsampling



Introduction

- Super-Resolution
 - Using multiple images



First image: Unger, Pock, Werlberger, Bischof 2010.

Second image: Goldlücke, Aubry, Kolev, Cremers 2014.

Introduction

- Optical Flow Estimation
 - Estimate the movement in the images



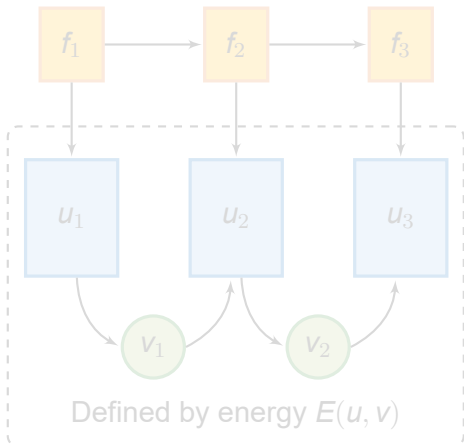


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Energy Minimization Approach

Input Images:



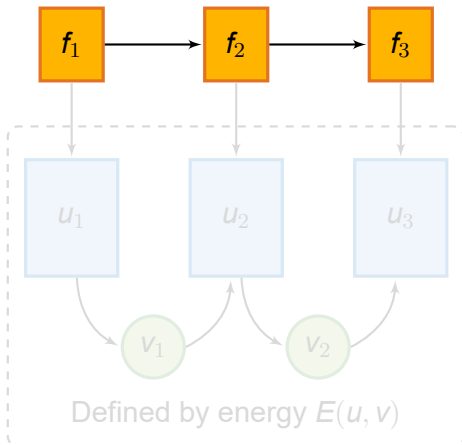
Alternating Optimization:

$$v^{k+1} \leftarrow \underset{v}{\operatorname{argmin}} E(u^k, v)$$

$$u^{k+1} \leftarrow \underset{u}{\operatorname{argmin}} E(u, v^{k+1})$$

Energy Minimization Approach

Input Images:



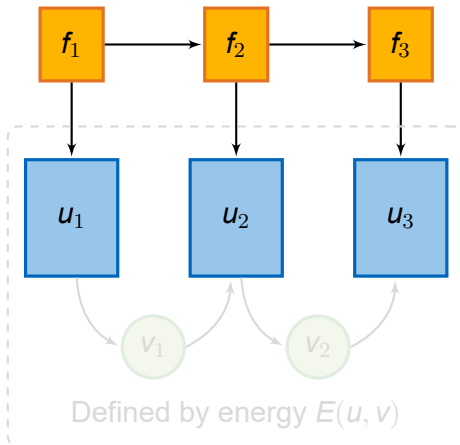
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Energy Minimization Approach

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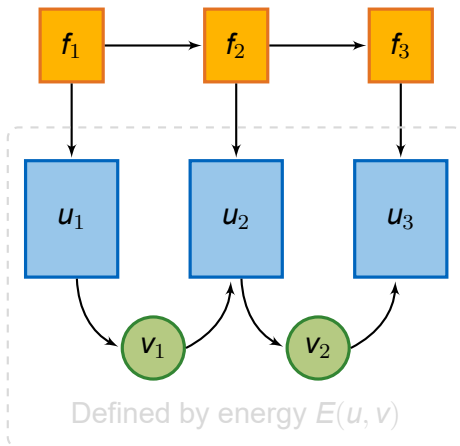
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Energy Minimization Approach

Input Images:



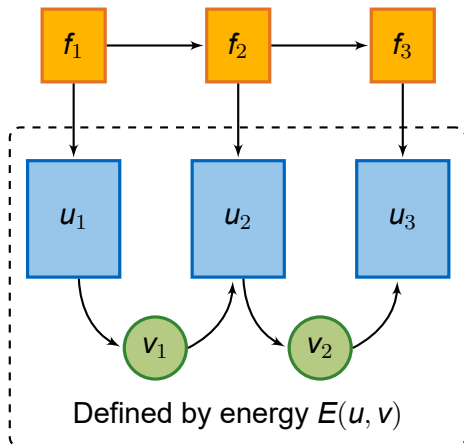
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Energy Minimization Approach

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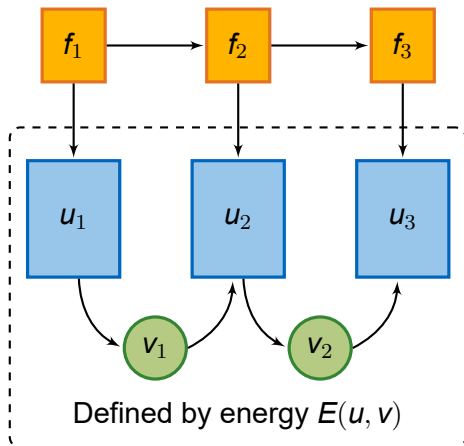
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Flow Field Energy

- Optical Flow Constraint: $u_i(x) \stackrel{!}{=} u_{i+1}(x + v_i(x))$
→ minimize $\|u_t - \nabla u^T \cdot v\|_1$
- Total Variation: $TV(v)$

$$E_{flow}(v) = \gamma \|u_t - \nabla u^T \cdot v\|_1 + TV(v)$$

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Super-Resolution Energy

- Super-Resolution

→ minimize: $\|Au - f\|_1$

- Optical Flow Constraint

→ minimize: $\|u_t - \nabla u^T \cdot v\|_1$

- Total Variation: $TV(u)$

$$E_{super}(u) = \alpha \|Au - f\|_1 + \beta TV(u) + \gamma \|u_t - \nabla u^T \cdot v\|_1$$

Super-Resolution Energy

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Super-Resolution Energy

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Total Energy

$$E_{\text{flow}}(v) = \gamma \|u_t - \nabla u^T \cdot v\|_1 + TV(v)$$

+

$$E_{\text{super}}(u) = \alpha \|Au - f\|_1 + \beta TV(u) + \gamma \|u_t - \nabla u^T \cdot v\|_1$$

↓

$$E(u, v) = \underbrace{\alpha \|Au - f\|_1 + \beta TV(u)}_{\text{Super-Resolution}} + \underbrace{TV(v)}_{\text{Flow}} + \underbrace{\gamma \|u_t - \nabla u^T \cdot v\|_1}_{\text{Coupling}}$$

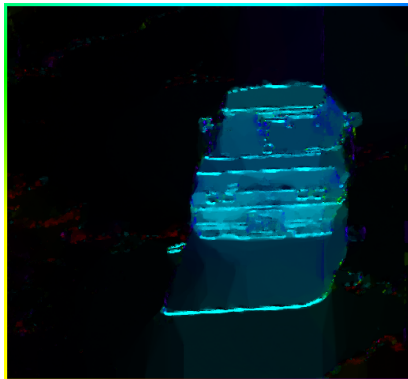


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Results

- Optical Flow Estimation
 - Live Demo ...



Results

■ Super-Resolution with Optical Flow Estimation

■ Input:



Input 1



Input 2

Results

- Super-Resolution with Optical Flow Estimation
 - Result:



Super-Resolution 1



Input 1

Results

- Super-Resolution with Optical Flow Estimation
 - Result:



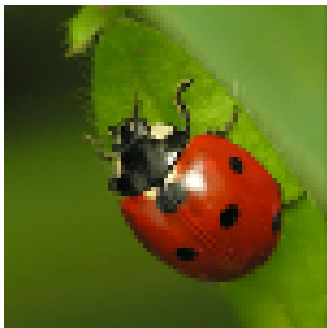
Super-Resolution 1



Input 1

Results

- Super-Resolution with Optical Flow Estimation
 - Input:



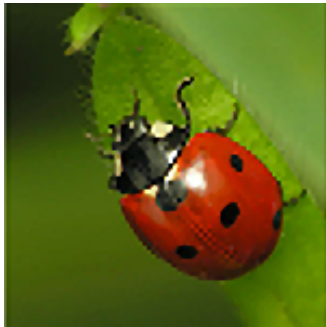
Downscaled



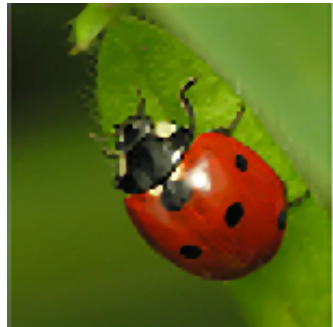
Original

Results

- Super-Resolution with Optical Flow Estimation
 - Result:



2 Input Images



4 Input Images



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Conclusion and Future Work

- Arbitrary scaling
- Optical flow estimation for movements > 1 Pixel



End

Thank you for your
attention!