

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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4 Future Work





Introduction

- Super-Resolution
 - **Example Pictures**



Introduction

- Optical Flow Estimation
 - Example Pictures



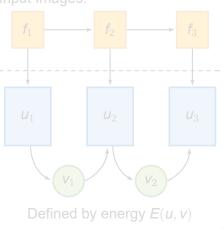
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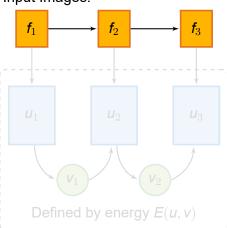






$$v^{k+1} \leftarrow \underset{v}{\operatorname{argmin}} \ E(u^k, v)$$
$$u^{k+1} \leftarrow \underset{u}{\operatorname{argmin}} \ E(u, v^{k+1})$$

Input Images:

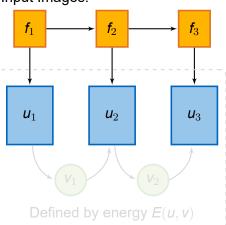


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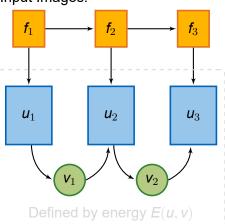
Alternating Optimization:

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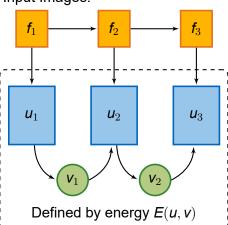
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Input Images:

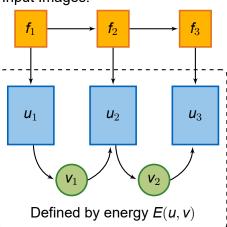


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Input Images:



Alternating Optimization:

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

- Optical Flow Constraint: $u_i(x) \stackrel{!}{=} u_{i+1}(x + v_i(x))$

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

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

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- Total Variation: *TV(v)*

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- Total Variation: *TV(v)*

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





- Super-Resolution
- \blacksquare 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
 - \rightarrow minimize: $||Au f||_1$
- \blacksquare Total Variation: TV(u)

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



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 - \rightarrow minimize: $||Au f||_1$
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 - \rightarrow minimize: $||u_t \nabla u^T \cdot v||_1$
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- Super-Resolution
 - \rightarrow minimize: $||Au f||_1$
- Optical Flow Contraint
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- Super-Resolution
 - \rightarrow minimize: $||Au f||_1$
- Optical Flow Contraint
 - \rightarrow minimize: $||u_t \nabla u^T \cdot v||_1$
- Total Variation: TV(u)

$$oxed{\mathsf{E}_{\mathsf{super}}(u) = lpha ||\mathsf{A} u - f||_1 + eta \mathsf{TV}(u) + \gamma ||u_t -
abla u^\mathsf{T} \cdot \mathsf{v}||_1}$$



Total Energy

$$\left[E_{flow}(\mathbf{v}) = \gamma ||\mathbf{u}_t - \nabla \mathbf{u}^T \cdot \mathbf{v}||_1 + TV(\mathbf{v}) \right]$$

$$oxed{E_{super}(u) = lpha ||Au - f||_1 + eta TV(u) + \gamma ||u_t -
abla u^T \cdot v||_1}$$

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



Computer Vision Group



Outline

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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





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Conclusion and Future Work

- More than two input images
- Arbitrary scaling
- Optical flow estimation for movements > 1 Pixel



End

Thank you for your attention!