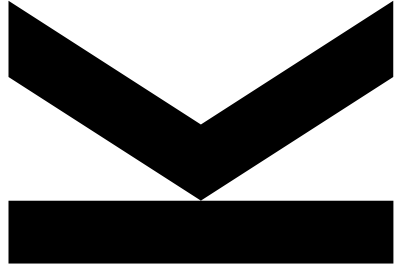


**JOHANNES KEPLER  
UNIVERSITY LINZ**

# INITIAL IDEAS AND FIRST TESTS



## Team B0:

Philipp Eberstaller

Dominik Heindl

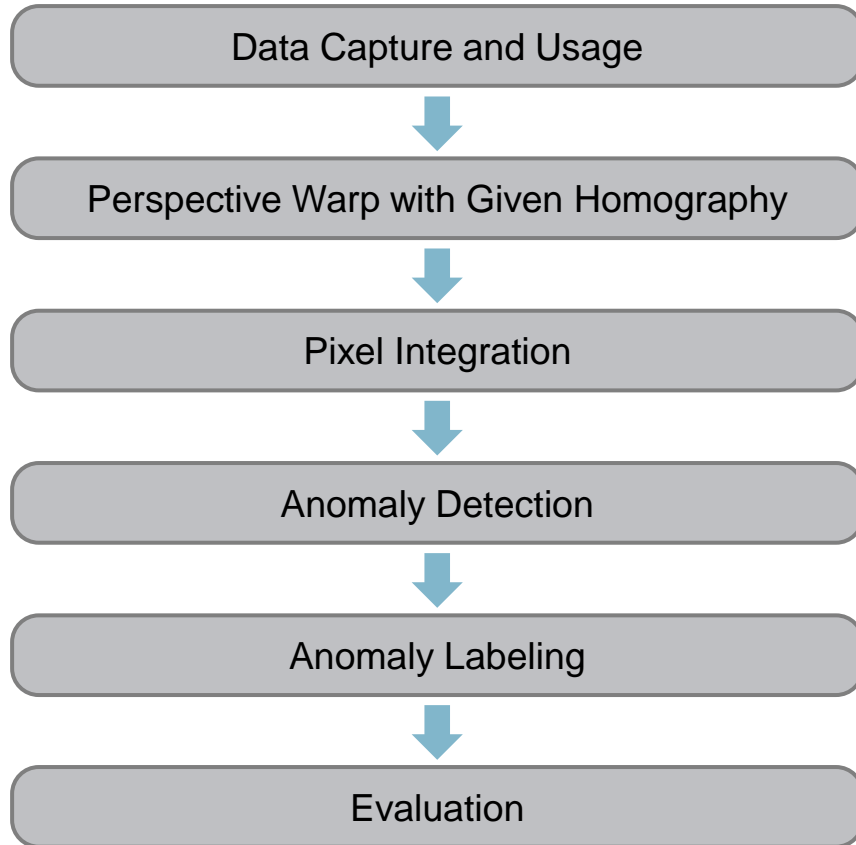
Carson Wittwer

# CONTENTS

- Implementation of Given Solution
- Idea Testing

# IMPLEMENTATION OF GIVEN SOLUTION

## Given Algorithm



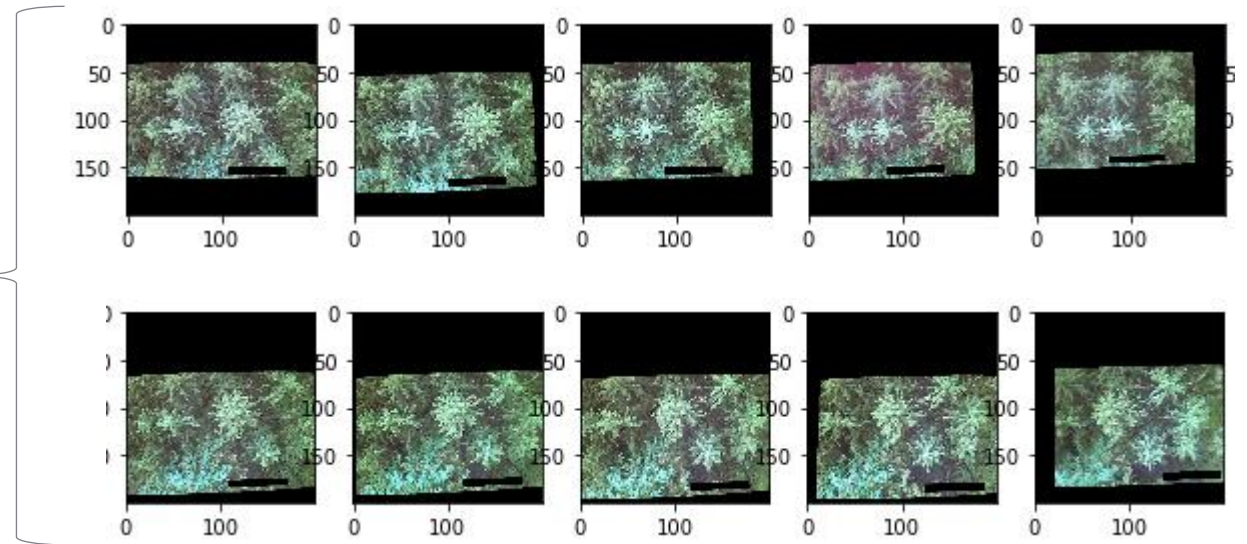
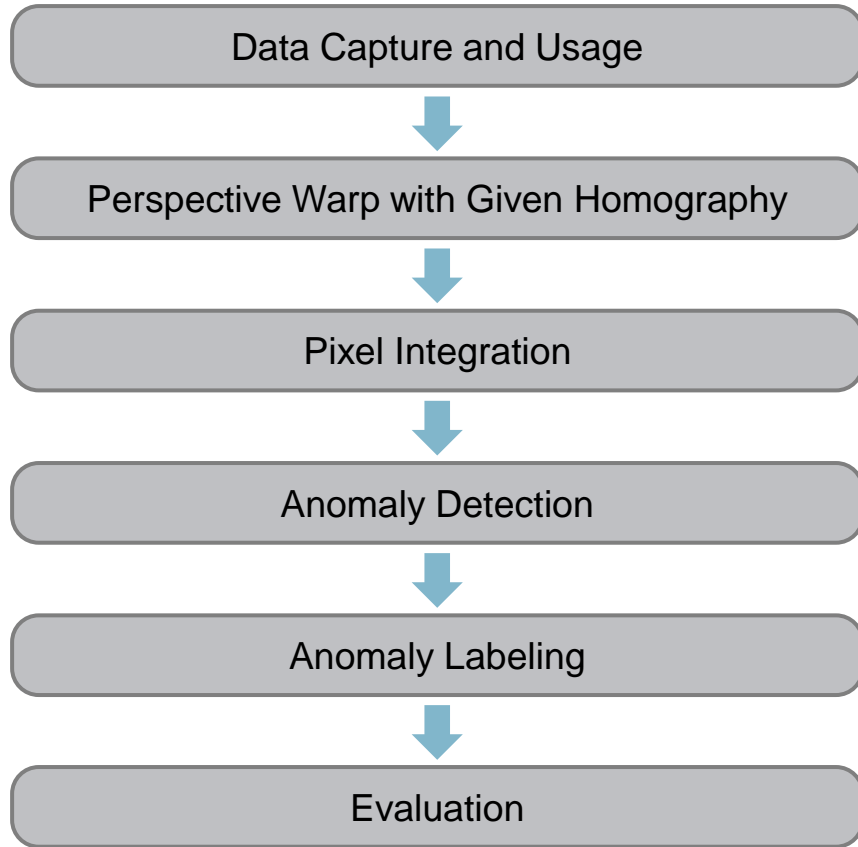
Most of our recent time has surrounded implementing the given solution.

Opportunity to:

- Implement basic functionality e.g. data loading, ...
- Understand the structure and flow of the algorithm
- Understand more specifics to what individual functions are doing
- Identify opportunities for improvement
- Identify where our ideas can fit in

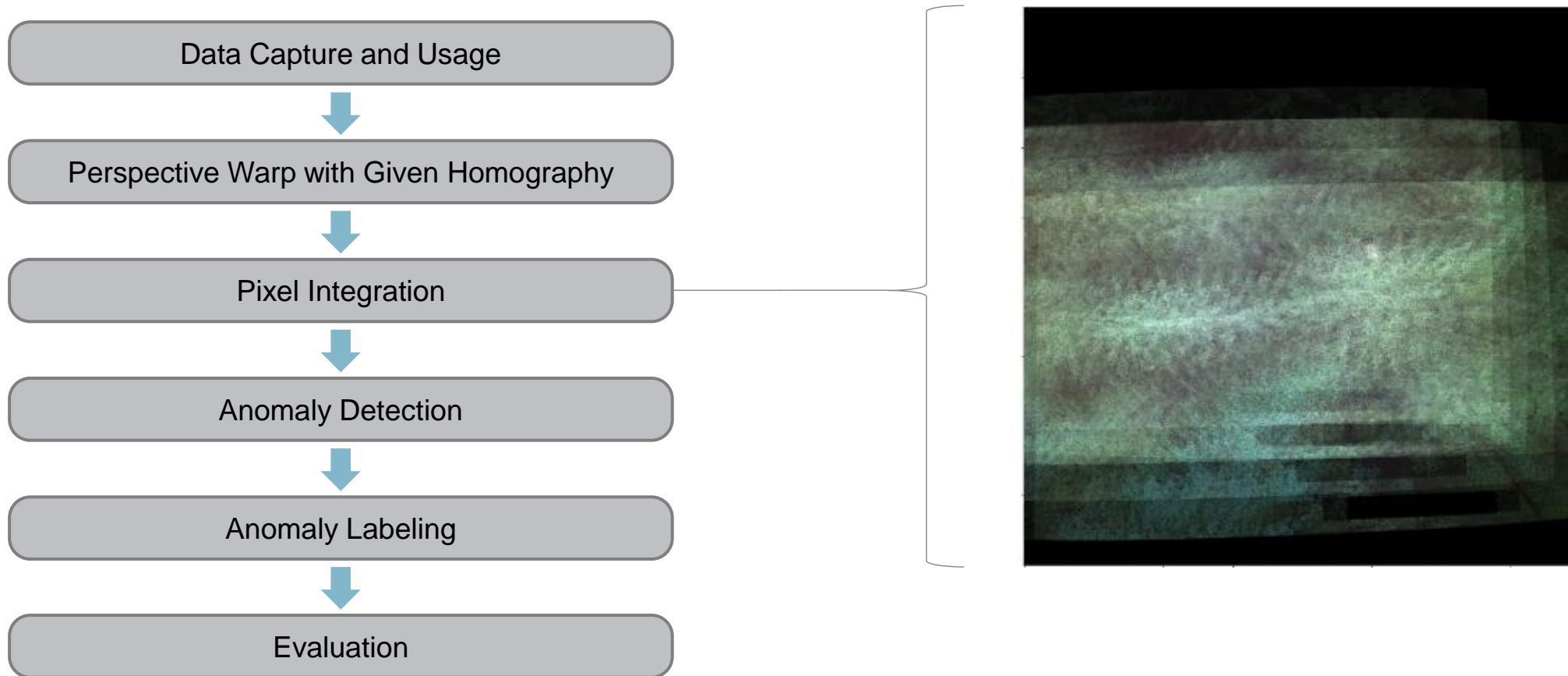
# IMPLEMENTATION OF GIVEN SOLUTION

## Given Algorithm



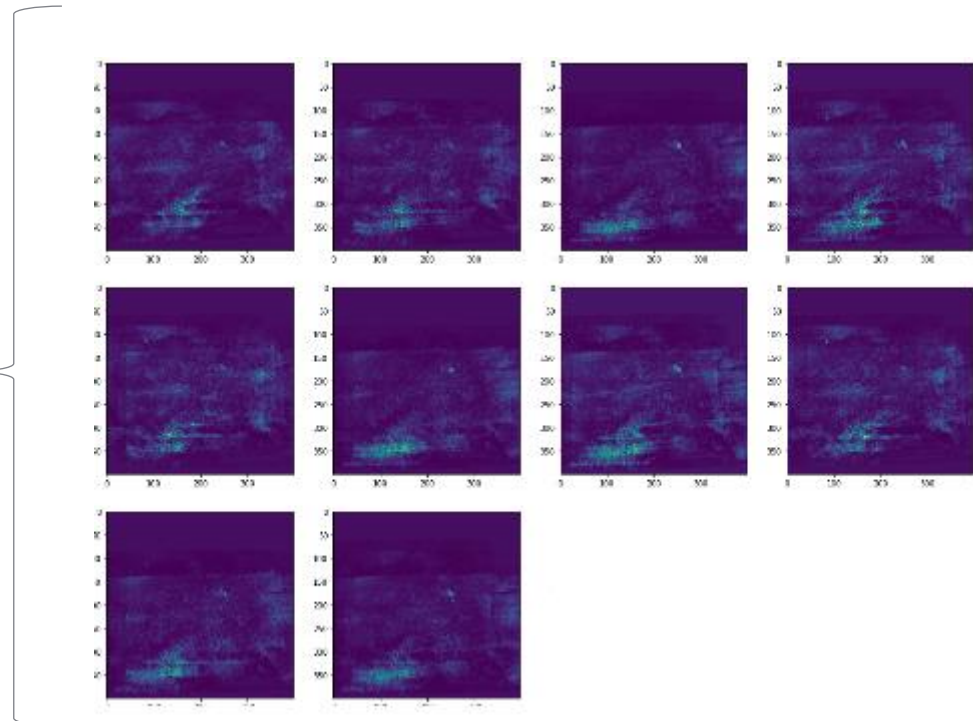
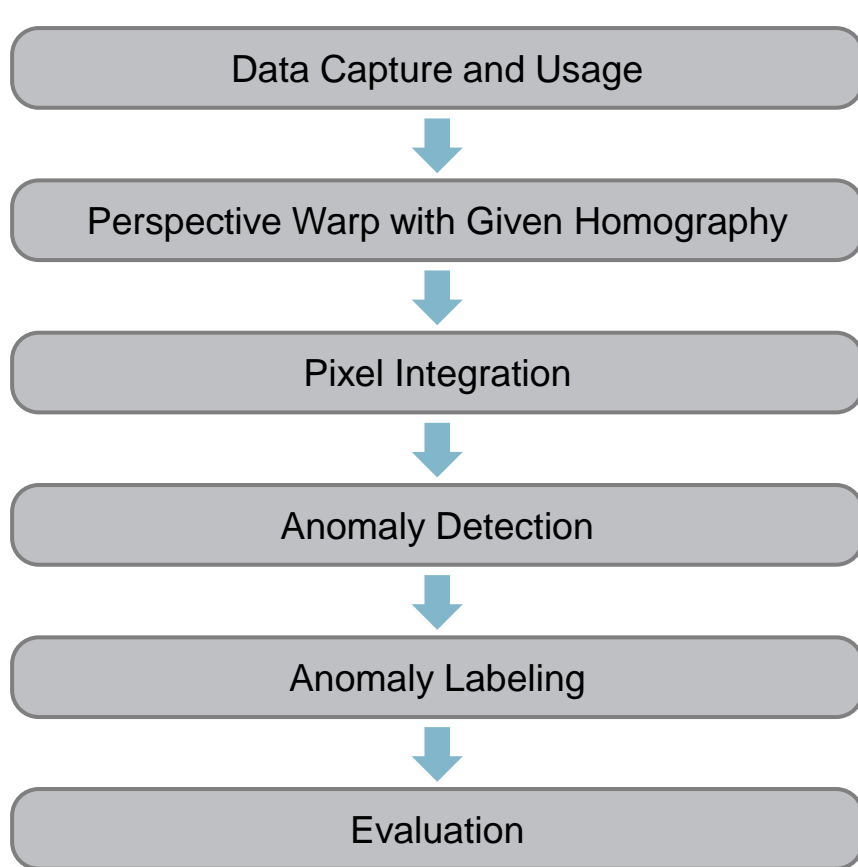
# IMPLEMENTATION OF GIVEN SOLUTION

## Given Algorithm



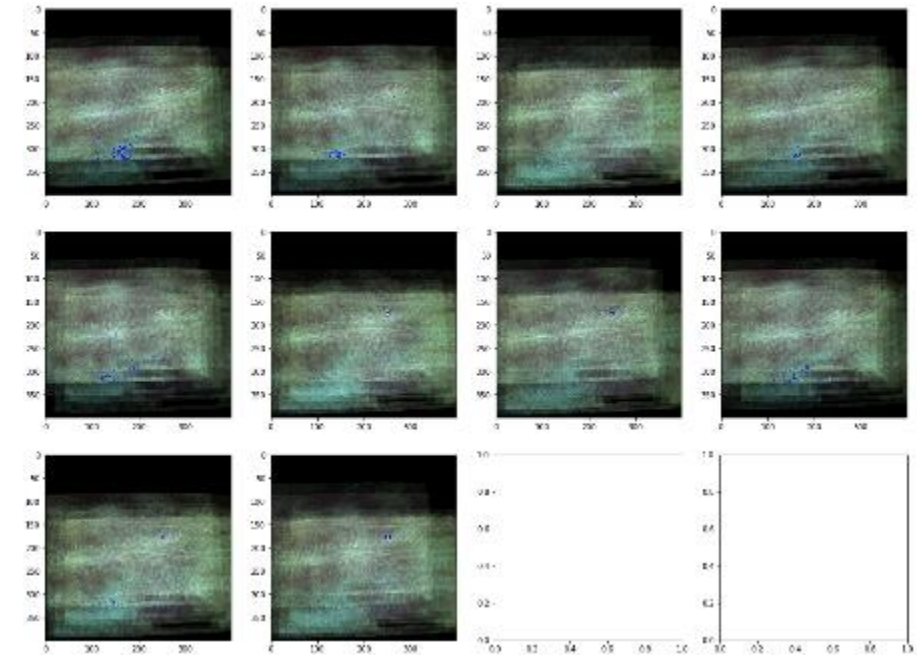
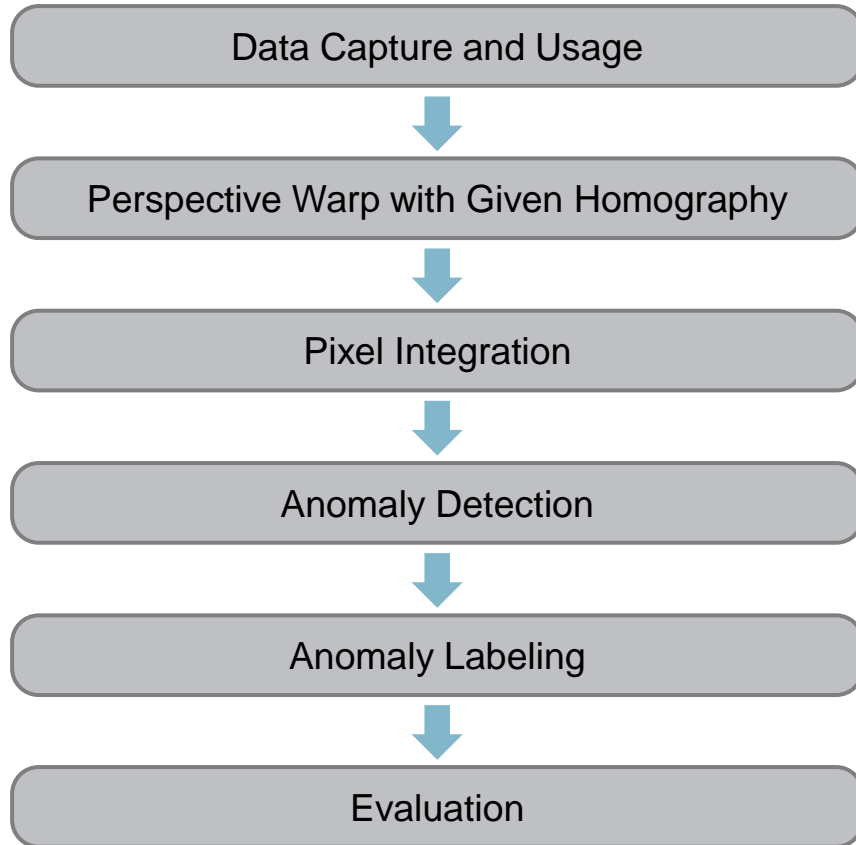
# IMPLEMENTATION OF GIVEN SOLUTION

## Given Algorithm



# IMPLEMENTATION OF GIVEN SOLUTION

## Given Algorithm



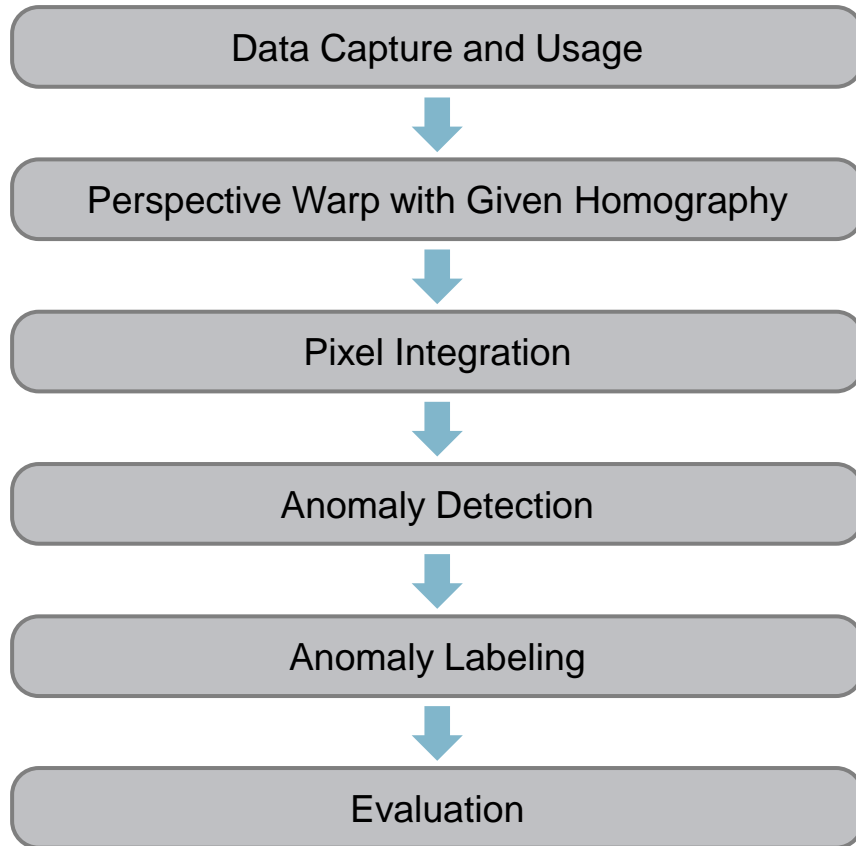


# CONTENTS

- Implementation of Given Solution
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# IDEA TESTING

## Given Algorithm



## Pseudocode for Pixel Labeling Via Energy Minimization

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**Algorithm** Labeling-based Synthetic Aperture Imaging

---

*Warp using homography using Eq.(1);*

**For** depth range  $l = 0$  **to**  $k$

*Shift each reference plane camera image to depth  $l$  using Eq.(3);*

**For** each pixel  $m$  of each camera view

*Compute  $E_m$  and set  $f_m$  using Eq.(9,13,15);*

**End**

*Shift back each camera image to the reference plane using Eq.(16);*

**End**

**For** camera view  $i = 1$  **to**  $N$

*Merge each labeled camera view on the reference plane together using Eq.(17);*

**End**

*Shift the labeled camera view from the reference plane to the desired focal plane using Eq.(3).*

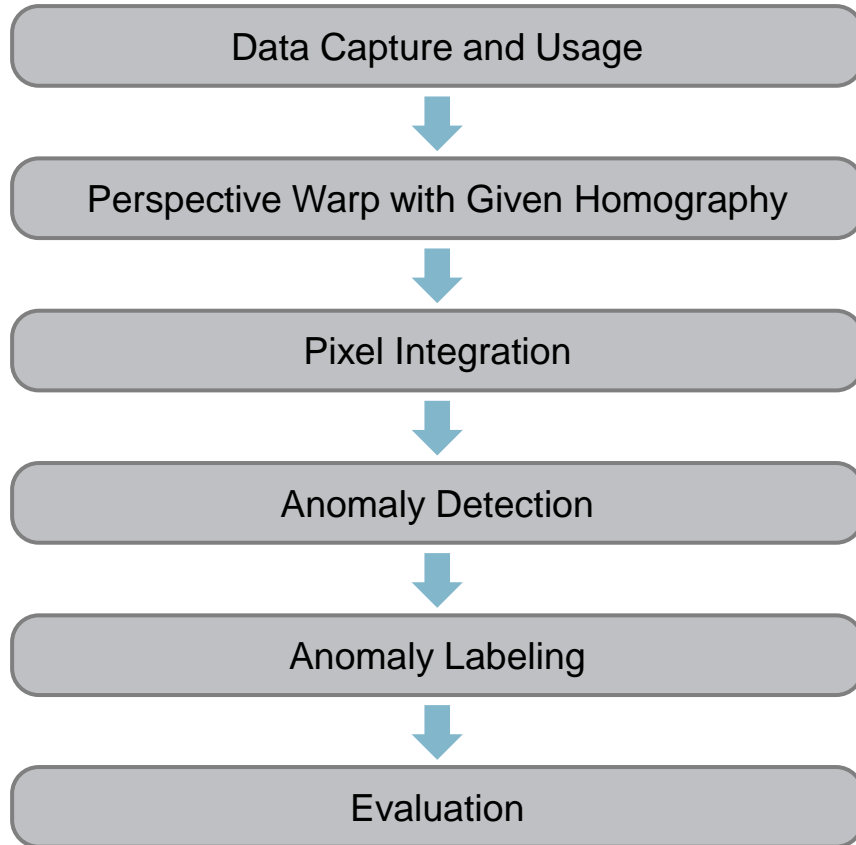
*Adding the images together without the labeled occlusion parts using Eq.(18);*

*Get the labeling-based synthetic aperture image on the desired focal plane;*

---

# IDEA TESTING

## Given Algorithm



## Pseudocode for Part of Detecting Anomalies in Noise Signal

---

### **Algorithm 1** Computation of the unstructured residual

---

**Require:** Multichannel Image  $u$ ,  $n$  the number of nearest neighbors

**Ensure:** Model  $\hat{u}$  of  $u$  based on  $\mathcal{D}$ , residual  $r(u) = \hat{u} - u$ .

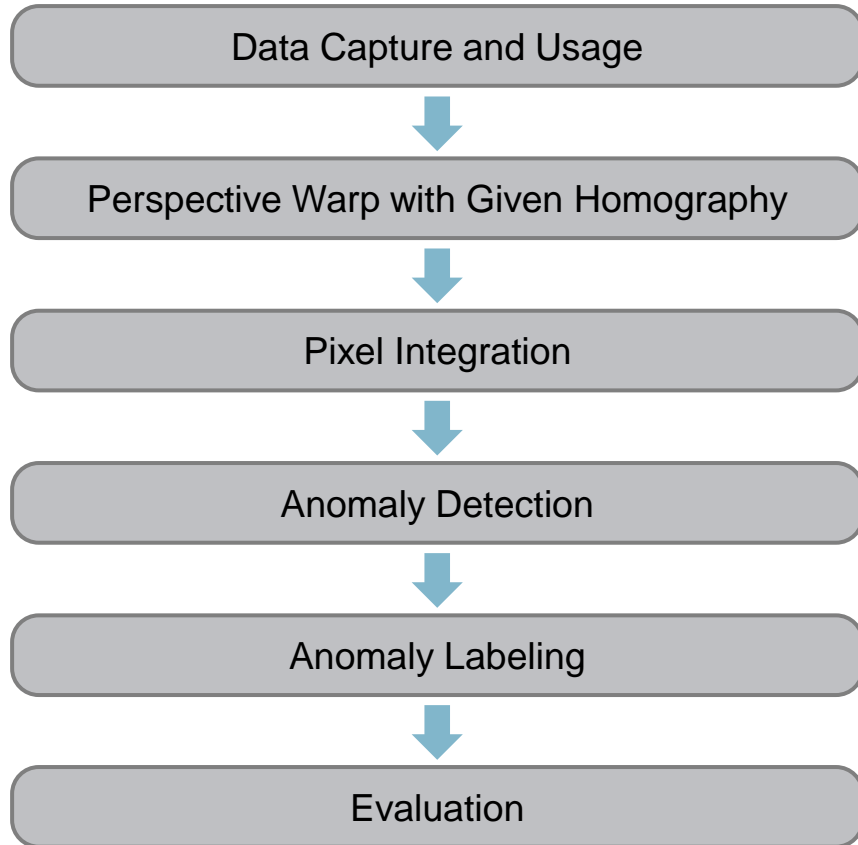
- 1: **for all** Multichannel patch  $P$  of  $u$  **do**
- 2:   Compute  $n$  near.neigh.  $\{P_i\}$  of  $P$  (**outside square region**).
- 3:   Reconstruct the patch (using (1))
- 4: **for all** pixels  $j$  in  $u$  **do**
- 5:    $\hat{u}(j) = \frac{\sum_{i \in \{s | j \in W_s, s \in [1, N]\}} \hat{P}_i(j)}{\#\{s | j \in W_s, s \in [1, N]\}}$

**Notation convention.**  $W_s$  : set of pixels in the patch centered at  $s$ .  
 $\hat{P}_i(j)$  : value at pixel  $j$  of the reconstructed patch centered at  $i$ .

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# IDEA TESTING

## Given Algorithm



Distinguish between anomalies → detect people

- use temporal data
- Check if anomalies are „moving“
  - Compare the location on different frames
- Stationary anomalies more likely to be non-human

**QUESTIONS?**