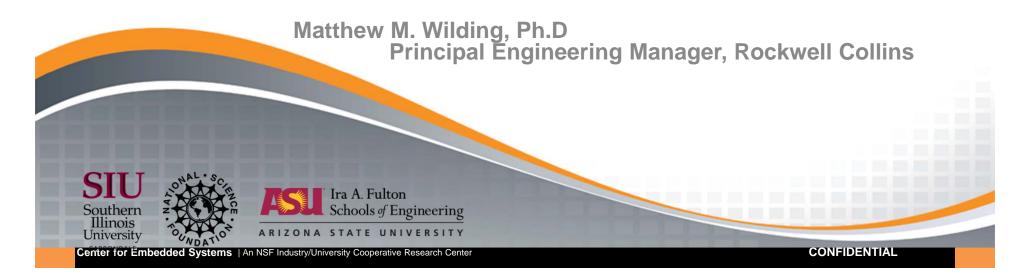


Registration and Fusion of EVS and SVS Runway Images for Embedded Systems

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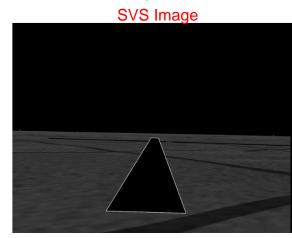


Project Overview and Description

Project Description

- The precise detection of runways is crucial for safely landing aircrafts because more than half of the accidents occur during the final approach and landing.
- The runway detection methodology being developed exploits information from enhanced vision system (EVS) and synthetic vision system (SVS) image frames of the runways.
- The goal is to generate image frames that contain enhanced runway and surrounding information by fusing the EVS and SVS frames.
- The resulting image frames can be incorporated into head-up displays (HUDs) to assist the pilot in landing the aircraft safely.





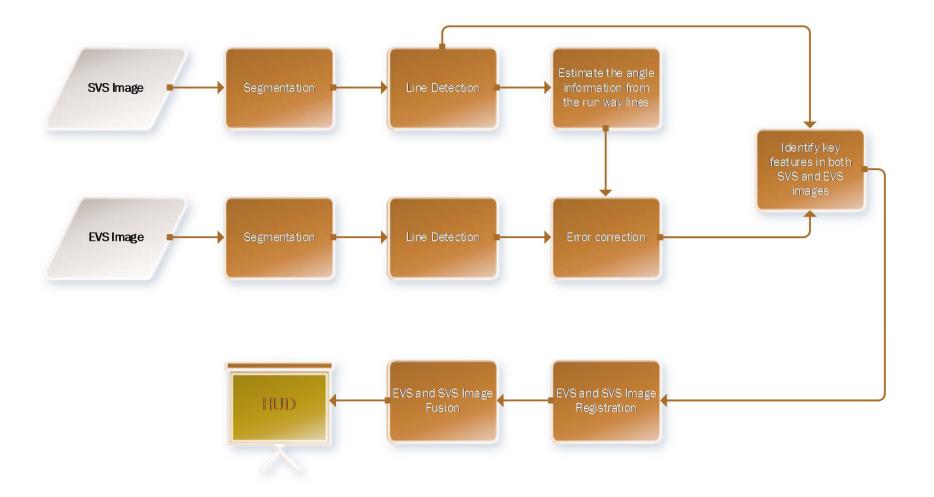
Approach

Novelty

- 1. Fusion of EVS and SVS images
- 2. Novel algorithms to:
 - (a) Extract features from EVS and SVS images
 - (b) Register EVS and SVS runway images
 - (c) Fuse the registered images so that the information from both images can be displayed optimally
 - (d) Simulate adverse weather condition images to objectively evaluate performance
- 3. Embed algorithms into multi-core processing environments for real-time applications

Benefits to member companies

- 1. Lead to the development of novel landing heads-up display systems
- 2. Will also be applicable to a vast range of other problems at Rockwell Collins involving image registration and image fusion.



Laplacian Line Detection Masks



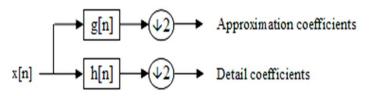
2	-1	-1
-1	2	-1
-1	-1	2

-1	2	-1
-1	2	-1
-1	2	-1

-1	-1	2
-1	2	-1
2	-1	-1

DWT

$$y_{low}[n] = (x * g)[n] = \sum_{k=-\infty}^{\infty} x[k] \cdot g[n-k]$$
$$y_{high}[n] = (x * h)[n] = \sum_{k=-\infty}^{\infty} x[k] \cdot h[n-k]$$



Fusion Objective

$$f_{ES}(x,y) = Fusion (f_E(x,y), f_S(x,y))$$

Such that the correlation

$$[f_{ES}(x,y) \otimes f_{E}(x,y)] + [f_{ES}(x,y) \otimes f_{S}(x,y)]$$

Is maximized

Rule 1: Maximum Selection Criterion

 $LL_{ES} = \max(LL_S, LL_E)$ $LH_{ES} = \max(LH_S, LH_E)$ $HL_{ES} = \max(HL_S, HL_E)$ $HH_{ES} = \max(HH_S, HH_E)$

Rule 2: Average Selection Criterion

 $LL_{ES} = ave(LL_S, LL_E)$ $LH_{ES} = ave(LH_S, LH_E)$ $HL_{ES} = ave(HL_S, HL_E)$ $HH_{ES} = ave(HH_S, HH_E)$

Rule 3: Maximum Selection for high frequency sub-bands (HL, LH, HH) and average selection for low frequency sub-band (LL)

 $LL_{ES} = \max(LL_S, LL_E)$ $LH_{ES} = \max(LH_S, LH_E)$ $HL_{ES} = \max(HL_S, HL_E)$ $HH_{ES} = \max(HH_S, HH_E)$

The fused image is obtained by taking the inverse wavelet transform after application of the selection rule.

Turbulence Model

$$H(u,v) = e^{-k(u^2+v^2)^{5/6}}$$

Weiner Filter

$$\widehat{F}(u,v) = \left[\frac{1}{H(u,v)} \frac{|H(u,v)|^2}{|H(u,v)|^2 + k}\right] G(u,v)$$

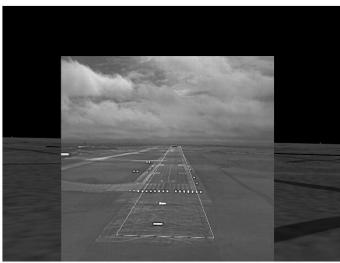
Objective evaluations

- Root Mean Square Error (RMSE) for registration
- Correlation Coefficient for fusion

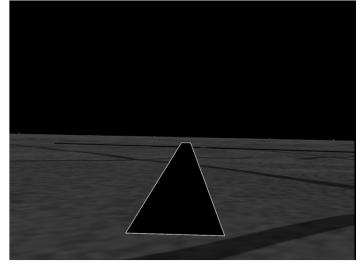
Results



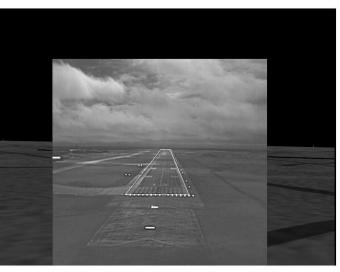
EVS Image



Fusion



SVS Image



Registration and Fusion

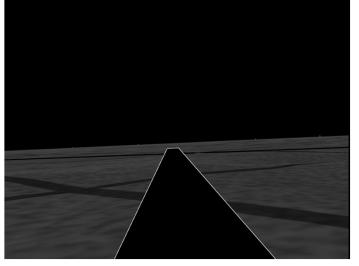
Results



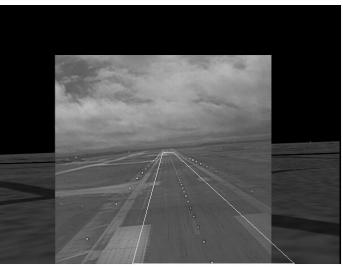
EVS Image







SVS Image

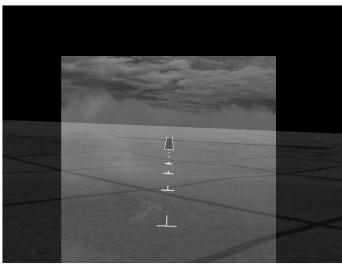


Registration and Fusion

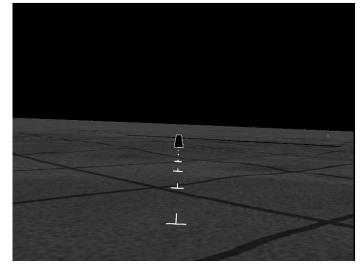
Fusion Results



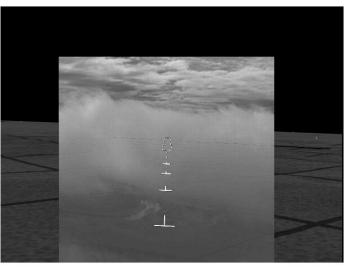
EVS Image



DWT Fusion (average)



SVS Image

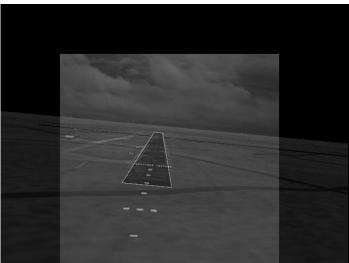


DWT Fusion (maximum)

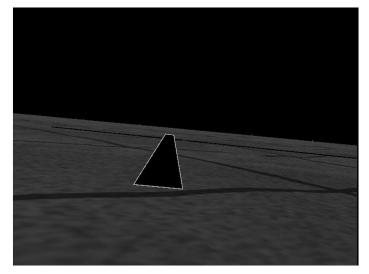
Fusion Results



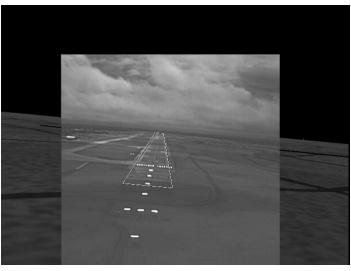
EVS Image







SVS Image



DWT Fusion (maximum)

Fusion Results



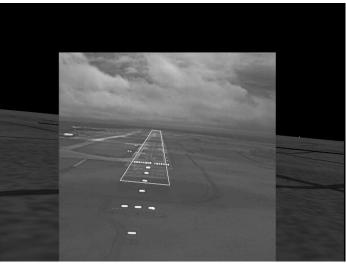
Averaging







DWT Fusion (Mixed)



GHM Fusion (maximum)

Turbulence



Original



Low Turbulence



Mild Turbulence

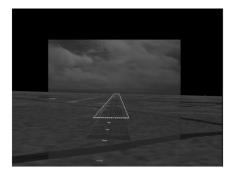


Severe Turbulence

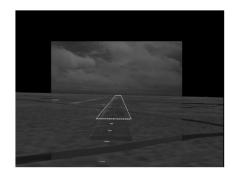
Registration & Fusion in Turbulence



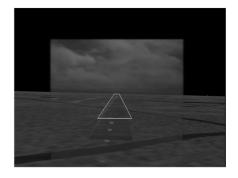
No Turbulence



Low Turbulence



Mild Turbulence



Severe Turbulence