



OBJECT TRACKING IN VIDEO AND LOCALIZATION

By

Somaia Mohamed Mahmoud Mohamed Mohamed

A Thesis Submitted to the Faculty of Engineering at Cairo University in Partial Fulfillment of the Requirements for the Degree of DOCTOR OF PHILOSOPHY

in

Electronics and Communications Engineering

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Title of Thesis:

Object Tracking and Navigation Based on Computer Vision

Key Words:

Computer Vision, Object tracking, SIFT, Localization, Visual Odometry

Summary:

This work proposes novel modify on algorithm for two different applications in computer vision namely localization based on vision and object tracking in video. The proposed work in this thesis is intended to improve the conventional system for object tracking through using wavelet transform and morphological operations. The new algorithm succeeds in removing the noise and smooth the image, as well as scale invariant feature transform (SIFT) to increase the number of features which, in turn, improves the matching stage. The proposed procedure has been applied on many consequent frames, under different circumstances. The implementation results show high accuracy and performance, compared with the other conventional procedures such as Wavelet-based object tracking algorithms, and SIFT-based object tracking algorithms.

On the other hand, the proposed navigation algorithm determines the location of the moving objects by using a computer vision system. A single camera system is used to select suitable features and a stereo camera system to obtain the locations of the object. This significantly improves the overall accuracy, compared with the conventional systems that mainly depend on maps to determine the location of the object by comparing the features of the image with the features of the map. The implementation results of the proposed object localization show high performance that achieved 10 meters' accuracy on the trajectory with a length of 165 meters.

Insert photo here

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Dedication

To My Beloved Family

My Parents,

My Brother,

and My sister

(Thank You So Much)

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Nomenclature

2-D Two-Dimensions

3-D Three-Dimensions

CDT Correct Detected Track
 Coif1 Coiflets wavelet base 1
 Coif2 Coiflets wavelet base 2
 CUPT Coordinate Update

DB2 Daubechies wavelet base 2
DB3 Daubechies wavelet base 3
DB4 Daubechies wavelet base 4
DB5 Daubechies wavelet base 5

DR Detection Rate

DWT Discrete Wavelet Transform

FAR
False Alarm Rate
FAT
False Alarm Track
FN
False Negative
FP
False Positive

GPS Global Positioning System

GT Ground Truth

HH High High subbands of Wavelet transformHL High Low subbands wavelet transform

HOG Histogram of Oriented Gradients

HSV (Hue, Saturation, Value)

ICA Independent Component Analysis
IDWT Inverse Discrete Wavelet Transform

IMU Inertial Measurement UnitINS Inertial Navigation Sensor

JPEG Joint Photographic Experts Group

LBP Local Binary Patterns

LH Low High sub band of Wavelet transform

LIDAR Light Detection And Ranging

LL Low Low subbands of Wavelet transform

PCA Principle Component Analysis
RANSAC Random sample consensus

RDWT Redundant Discrete Wavelet transform

SFM Structure From Motion

SIFT Scale Invariant Features Transform
SLAM Simultaneous Localization and Mapping

ST System Truth

SURF Speed Up Robust FeaturesSVD Singular Value Decomposition

Sym2 Symlets wavelet base 2 Sym4 Symlets wavelet base 4 **TDF** Track Detection Failure

TF total number of frames in the video sequence
TG total number of frames for the ground truth objects

TN True Negative TP True Positive

TRDR Tracker Detection Rate
VBN Vision Based Navigation

VO Visual Odometry

V-SLAM Visual Simultaneous Localization and Mapping

WT Wavelet Transform

List of symbols

Alph_c Skew of camera

 $\begin{array}{lll} cc & principle \ point \ of \ camera \\ Fc & Focal \ Length \ of \ camera \\ Kc & distortion \ of \ camera \\ I_t(x,y) & Image \ in \ Spatial \ domain \\ B_t & Background \ model \\ T & Experimental \ Threshold \end{array}$

 μ_d Mean

 λ_d : Standard deviation

T_s Foreground experimental threshold

L scale space of an image

G scale Gaussian

D Difference of Gaussian
 m gradient magnitude
 Θ Gradient orientation
 H Hessian matrix

 L_{xx} convolution of the second order derivative of Gaussian

with an image in x direction

L_{xx} convolution of the second order derivative of Gaussian

with an image in y direction

λ Eigen values

R The measure of corner response

tr Trace of matrix
Det determinant of matrix

A spatial overlap between ground truth and system truth
O Binary variable between ground truth and system truth

Tov Threshold

TR_{ov} arbitrary threshold

d Distance