

tidy.txt

Column number =>

1 2 3 4 5 6 7 8 9 10 11 12

Variable or Column
Name =>

SubjectID
ActivityID
BodyAcc-mean-X
BodyAcc-mean-Y
BodyAcc-mean-Z
BodyAcc-std-X
BodyAcc-std-Y
BodyAcc-std-Z
GravityAcc-mean-X
GravityAcc-mean-Y
GravityAcc-mean-Z
GravityAcc-std-X
GravityAcc-std-Y
GravityAcc-std-Z
BodyA
P

filename: tidy

format: text comma delimited, 264 kb

See run_Analysis.R program for ETL details.

Dataset documentation available at:http://archive.ics.uci.edu/ml/dataset_documentation.html

Data source:<https://d396qusza40orc.cloudfront.net/getdata>

Date created: 11/18/2014

Study Design

The Human Activity Recognition using smart activities of daily living (adl), e.g., walking u Linear acceration data of these ads were c samsumg galaxy SII smartphone. 3 axial ang embedded gyroscope of the samsung phon These sensor signals (triaxial and bodily acc and gyroscope were refined by apply noise to refine and sharpen the signal, allowing g These signals were further calculated into v The dataset is randmonly partitioned into tr

contents: A summary (mean) of select activity measur mean of the variable for each subject's acti measured by samsung II smartphone readir

The summary observations (means of selected
legacy (unzipped) folders UCI HAR Dataset\
Accelerometer data are in standard gravity
Angular velocity data (from the gyroscope)
SubjectID and ActivityID are integers while
see codebook tab for descriptions of variables

citation/source information:

Davide Anguita, Alessandro Ghio, Luca Oneto
Human Activity Recognition on Smartphones
International Workshop of Ambient Assisted Living

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www.smartlab.ws

dataset file layout

13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34	
BodyAcc-std-Z	BodyAcc-std-X	BodyAcc-std-Y	BodyAccJerk-std-Z	BodyAccJerk-std-X	BodyAccJerk-std-Y	BodyGyro-std-Z	BodyGyro-std-X	BodyGyro-std-Y	BodyGyroJerk-std-Z	BodyGyroJerk-std-X	BodyGyroJerk-std-Y	BodyAccMag-mean	BodyAccMag-std	GravityAcc	GravityMag	GravityStd	GravityMean	GravityStd	GravityMean	GravityStd	GravityMean	GravityStd

<https://archive.uci.edu/ml/datasets/Human+Activity+Recognition+Using+Smartphones>
data%2Fprojectfiles%2FUCI%20HAR%20Dataset.zip

The database describes 30 subjects aged 19 - 48 performing
upstairs, walking downstairs, sitting, standing, laying.

Data was collected by the embedded accelerometer on each subject's

Angular velocity data of these sensors were collected by the

device.

(Acceleration and triaxial angular velocity) from the accelerometer
using low-pass filters.

Gravitational and body motion components to be separated.

Various features (total of 561) from their associated time and frequency domains.

Training (70% of all subjects) and test (30% of all subjects) partitions.

Features across 128 observations from 30 subjects over 5 activities.

Activity (walking, walking upstairs, walking downstairs, sitting standing, laying)

Features from accelerometer and gyroscope. All values are numeric.

t activity variable) are derived from original raw data contained in
(test or train)\Inertial Signals
units. Body acceleration signals are calculated by subtracting gravity from total acceleration
are in radians/second units.
all other variables are numbers with a decimal point, e.g., -0.2853077700. No missing values
file or column headings

to, Xavier Parra and Jorge L. Reyes-Ortiz.
as using a Multiclass Hardware-Friendly Support Vector Machine.
d Living (IWAAL 2012). Vitoria-Gasteiz, Spain. Dec 2012

essandro Ghio, Luca Oneto.
s Laboratory

3.

4.

t

[illegible]

atin

/values (NA).

[illegible]

79 80 81 82 83 84 85 86 87 88

ean
Mag-std
Mag-meanFreq
dyAccMean,gravity.
clerkMean.,gravityMean.
e.BodyGyroMean,gravityMean.
le.BodyGyroJerkMean,gravityMean.
angle.X,gravityMean.
angle.Y,gravityMean.
angle.Z,gravityMean.