

FEATURE EXTRACTION USING MATLAB IN < 30 MINUTES

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DEPARTMENT OF ELECTRICAL ENGINEERING



ORS Research Design & Data Analysis Lab Office of Research and Scholarship

OUTLINE

>INTRODUCTION

>DIFFERENT TYPES OF FEATURES

► DISCUSSION

OUTLINE

➤INTRODUCTION

>DIFFERENT TYPES OF FEATURES

► DISCUSSION

>Why do we need to derive features from data ?

- To test an hypothesis
- To classify different groups
- To predict outcomes







>Feature extraction is the process of transforming raw data into numerical features while preserving the information of the original data set.



>Feature extraction identifies most discriminating characteristics in signals.





OUTLINE

>INTRODUCTION

>DIFFERENT TYPES OF FEATURES

► DISCUSSION

► VARIABLE

>STATISTICAL FEATURES

>SPECTRAL FEATURES

>NONLINEAR FEATURES

≻VARIABLE			<pre>lm = fitlm(tbl.'MPG</pre>	<pre>lm = fitlm(tbl,'MPG~Weight+Acceleration')</pre>					
	fitli	n	lm =						
Weight	Acceleration	MPG							
			Linear regression model: MPG ~ 1 + Weight + Acceleration						
3504	12	18	Estimated Coefficie	ents:					
3693	11.5	15		Estimate	SE	tStat	pValue		
3436	11	18							
3433	12	16	(Intercept) Weight	45.155 -0.0082475	3.4659 0.00059836	13.028 -13.783	1.6266e-22 5.3165e-24		
3449	10.5	17	Acceleration	0.19694	0.14743	1.3359	0.18493		
			MPG =	a + b Weig	ght + c Acce	leration			

```
Number of observations: 94, Error degrees of freedom: 91
Root Mean Squared Error: 4.12
R-squared: 0.743, Adjusted R-Squared 0.738
```

VARIARIE	Command Window					Workspace			
VANIADLL	<pre>>> lm=fitlm(housing)</pre>								
						housing Im			
	lm =								
	Linear regression model:								
	median_house_value ~ [Linear formula with 9 terms in 8 predictors]								
aal Estata Data									
eal Estate Data	Estimated Coefficients:	Ratimata	A E						
		Estimate	SE	tStat	pvalue				
	(Intercept)	-3.5854e+06	62901	-57.001	0				
	(Intercept) longitude	-3.5854e+06 -42730	62901 717.09	-57.001 -59.588	0 0				
	(Intercept) longitude latitude	-3.5854e+06 -42730 -42510	62901 717.09 676.95	-57.001 -59.588 -62.796	0 0 0				
	(Intercept) longitude latitude housing_median_age	-3.5854e+06 -42730 -42510 1157.9	62901 717.09 676.95 43.389	-57.001 -59.588 -62.796 26.687	0 0 2.9463e-154				
	(Intercept) longitude latitude housing_median_age total_rooms	-3.5854e+06 -42730 -42510 1157.9 -8.2497	62901 717.09 676.95 43.389 0.79426	-57.001 -59.588 -62.796 26.687 -10.387	0 0 2.9463e-154 3.2948e-25				
	(Intercept) longitude latitude housing_median_age total_rooms total_bedrooms	-3.5854e+06 -42730 -42510 1157.9 -8.2497 113.82	62901 717.09 676.95 43.389 0.79426 6.9306	-57.001 -59.588 -62.796 26.687 -10.387 16.423	0 0 2.9463e-154 3.2948e-25 3.1889e-60				
	(Intercept) longitude latitude housing_median_age total_rooms total_bedrooms population	-3.5854e+06 -42730 -42510 1157.9 -8.2497 113.82 -38.386	62901 717.09 676.95 43.389 0.79426 6.9306 1.0841	-57.001 -59.588 -62.796 26.687 -10.387 16.423 -35.407	0 0 2.9463e-154 3.2948e-25 3.1889e-60 1.4597e-266				
	(Intercept) longitude latitude housing_median_age total_rooms total_bedrooms population households	-3.5854e+06 -42730 -42510 1157.9 -8.2497 113.82 -38.386 47.701	62901 717.09 676.95 43.389 0.79426 6.9306 1.0841 7.5466	-57.001 -59.588 -62.796 26.687 -10.387 16.423 -35.407 6.3209	0 0 2.9463e-154 3.2948e-25 3.1889e-60 1.4597e-266 2.6535e-10				

► VARIABLE Don't want to write the code? đ \times 🕜 Help 📥 🎃 👷 🞯 🗉 🛛 fitlm 🛛 🗶 🕇 Documentation Search Help Close ■ CONTENTS fitlm < Documentation Home Create linear regression model collapse all in page < Statistics and Machine Learning A Toolbox fitIm creates a LinearModel object. Once you create the object, you can see it in the workspace. You can see all the properties the object contains by clicking on it. You can create plots and do further diagnostic analysis by using < Regression methods such as plot, plotResiduals, and plotDiagnostics. For a full list of methods for LinearModel, see methods < Linear Regression < Multiple Linear Regression Syntax < Statistics and Machine Learning mdl = fitlm(tbl) Toolbox mdl = fitlm(tbl,modelspec) < Regression mdl = fitlm(X,y)< Linear Regression mdl = fitlm(X,y,modelspec) < Stepwise Regression mdl = fitlm(____, Name, Value) < Statistics and Machine Learning Toolbox < Functions Description example mdl = fitlm(tbl) returns a linear model fit to variables in the table or dataset array tbl. By default, fitlm takes the last variable as the response variable. fitlm ON THIS PAGE example mdl = fitlm(tbl,modelspec) returns a linear model of the type you specify in modelspec fit to variables in the table or dataset array tbl. Syntax example mdl = fitlm(X,y) returns a linear model of the responses y, fit to the data matrix X. Description example mdl = fitlm(X,y,modelspec) returns a linear model of the type you specify in modelspec for the responses y, fit to the data matrix X. Examples example Input Arguments mdl = fitlm(____, Name, Value) returns a linear model with additional options specified by one or more Name, Value pair arguments. Output Arguments For example, you can specify which variables are categorical, perform robust regression, or use observation weights More About Examples Tips Extended Capabilities ✓ Fit Linear Regression Using Data in Table See Also Load the sample data. **Open Script**





Kurtosis : *r* = 0.18 *p* = 0.03

Atluri, Salvatore and Indic (unpublished)



→ SPECTRAL FEATURES (a) 4000 9 2000 0 2000

wavelets(filename)



Indic P, Salvatore P, Maggini C, et al. (2011) Scaling Behavior of Human Locomotor Activity Amplitude: Association with Bipolar Disorder. PLOS ONE 2011 6(5): e20650.

>SPECTRAL FEATURES



Indic P, Murray G, Maggini C, et al. (2012) Multiscale Motility Amplitude Associated with Suicidal Thoughts in Major Depression. PLOS ONE 2012, 7: e38761







Bloch-Salisbury E, Indic P, Bednarek F, and Paydarfar D, J Appl Physiol., 2009, 107: 1017-1027



> SPECTRAL FEATURES

>NONLINEAR FEATURES



- Fluctuation Analysis
- Pattern Analysis
- Fractal Analysis
- Information Categorization Approach
- Power Law
- Entropy
- Dimension



Yang CC, Peng CK, Yien HW, Goldberger AL. Information categorization approach to literary authorship disputes. Physica A. 2003; 329:473-483.



Carreiro, S, Chintha KK, Shrestha S*, Chapman B, Smelson D, Indic P. Wearable sensor based detection of stress and craving in patients during treatment for substance use disorder: A mixed methods pilot study. Drug and Alcohol Dependence. 2020, 107929

Additional MATLAB Code

Physionet: https://physionet.org/

TISEAN: https://www.pks.mpg.de/~tisean/

SUMMARY

ORS Research Design & Data Analysis Lab

Office of Research and Scholarship





The University of Texas at **TYLER** Center for Health Informatics & Analytics **Ready to go features** 0 CHIA FEATURES f(x)dx**Statistical or** fast & efficient **Machine Learning Models** Feature Selection Feature Extraction Preprocessing Data

THANK YOU

Current Students:



Sloke Shrestha, UG



Mohammed Alenazi, Graduate



Pravitha Ramanand, PhD, Postdoc



Joshua Stapp, Graduate

Former Students:

Apurupa Amperayani (PhD Student, Arizona State University) Jonathan Wells (PhD Student, UT Austin) Pallavi Atluri Keerthi Chintha (Data Scientist, Wabtec Corporation) Selorm Darkey (Business Intelligent Analyst, Taylor Solutions)





THANK YOU

SBIR: RAE (Realize, Analyze, Engage) - A digital biomarker based detection and intervention system for stress and carvings during recovery from substance abuse disorders. *PIs: M. Reinhardt, S. Carreiro, P. Indic*

STARs Award

The University of Texas System *P. Indic (PI, UT Tyler)*

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Department of Veterans Affairs

Design of a wearable sensor system and associated algorithm to track suicidal ideation from movement variability and develop a novel objective marker of suicidal ideation and behavior risk in veterans. Clinical Science Research and Development Grant (approved for funding),

P. Indic (site PI, UT-Tyler)

E.G. Smith (Project PI, VA)

P. Salvatore (Investigator, Harvard University)



Design of a wearable biosensor sensor system with wireless network for the remote detection of life threatening events in neonates.

National Science Foundation Smart & Connected Health Grant

P. Indic (Lead PI, UT-Tyler)

D. Paydarfar (Co PI, UT-Austin)

H. Wang (Co PI, UMass Dartmouth)

Y. Kim (Co PI, UMass Dartmouth)



Pre-Vent

National Institute Of Health Grant *P. Indic (Analytical Core PI, UT-Tyler) N. Ambal (PI, Univ. of Alabama, Birmingham)*

Wearable system for the detection of addiction P. Indic (PI, UT-Tyler) M. Reinhart (PI, ContinueYou, LLC S. Carriero, (PI. Univ. of Mass. Med. School)

DISCUSSION