



### Numerical investigation on multiclass probabilistic classification of damage location in a plate structure

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#### taken from <u>www.ptclwg.com</u>



#### taken from <u>www.rbengineering.com</u>

## Solution

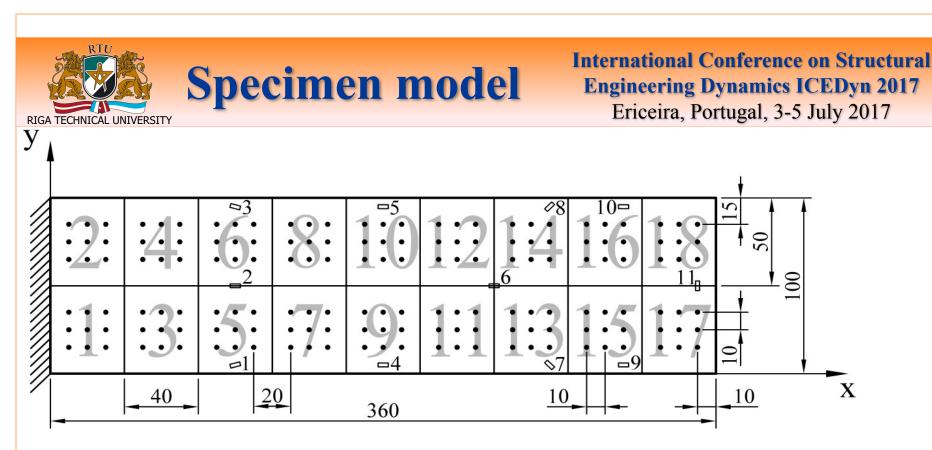
nondestructive structural health monitoring methods



# Damage localisation in thin composite structures based on machine learning algorithms

k- nearest neighbours

**Decision trees** 



- cantilevered CFRP plate (360 x 10 x 2.4 mm)
- Laminate lay-up [90/90/0/0/45/45/-45/-45/-45/45/0/90]<sub>s</sub>
- $E_x = 110 \text{ GPa}, E_y = 7 \text{ GPa}, G_{xy} = G_{yz} = 4.5 \text{ GPa}, v_{xy} = 0.33, \rho = 1560 \text{ kg/m}^3.$
- 11 strain sensors



### ANSYS model – 8-node shear deformable shell elements (72 x 20 elements)

**Damage** – an artificial mass with 5 % and 10 % fractions of plate's mass is placed at selected nodes of the plate. Additional mass is applied by using **MASS21 finite element**.

Modal analysis (block Lanczos method) to extract **4** eigenfrequencies and eigenmodes.



## Class labels

#### Plate is partitioned into 18 zones



Damage is applied to 9 points in each zone



RIGA TECHNICAL UNIVERSITY Damage localiza	atio	<b>International Conference on Structural</b> <b>Engineering Dynamics ICEDyn 2017</b> Ericeira, Portugal, 3-5 July 2017
Input strain values for each subzone	$\rightarrow$	162 subzones × 11 strain sensors
	$\rightarrow$	<i>k</i> -NN (define <i>k</i> and distance)
Build a classification model	$\rightarrow$	decision trees (define max number of splits)
Calculate resubstitution loss	$\rightarrow$	<i>k</i> -NN (update <i>k</i> and distance to yield minimum)
	$\rightarrow$	decision trees (update max number of splits to yield minimum)
Cross-validate the model	$\rightarrow$	<i>k</i> -NN and decision trees (update K to yield min cross-validation error)
	$\rightarrow$	Compute confusion matrix and ROC curve
Make prediction for future data	$\rightarrow$	Estimate posterior probabilities
Classify new unknown data in terms of	$\rightarrow$	Perform k-NN search
affiliation to any of 18 zones	$\rightarrow$	Build a decision tree

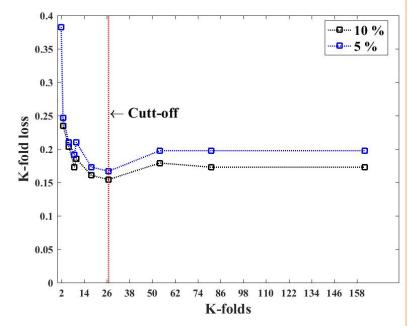
Make a decision regarding location of damage based on majority voting for 5 % and 10 % damage severities



# RESULTS

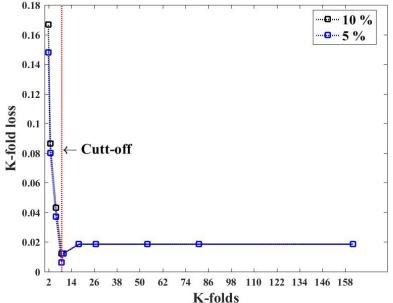


**Decision trees** 



Damage severity	10 %	5 %
Number of K-folds	27	27
K-fold loss (%)	15.43	16.67

k-NN

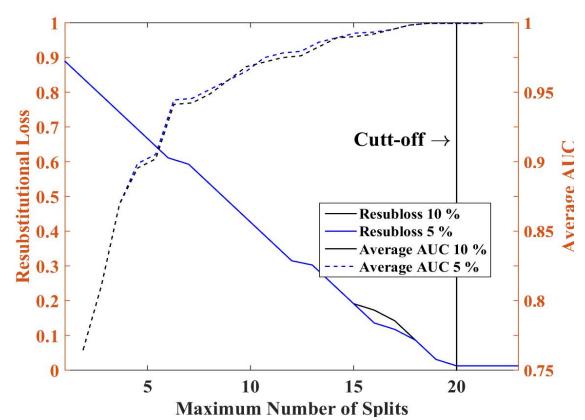


Damage severity	10 %	5 %
Number of K-folds	9	9
K-fold loss (%)	0.62	0.62



### **Resubstitution error**

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#### *k*-NN

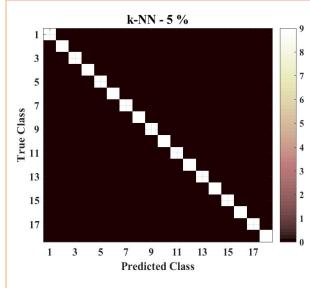
#### **Decision trees**

Damage severity	10 %	5 %
k	3	3
<b>Resubstitution loss (%)</b>	0	0

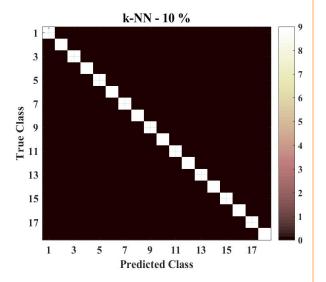
Damage severity	10 %	5 %
Maximum number of splits	3	3
Resubstitution loss (%)	1.23	1.23

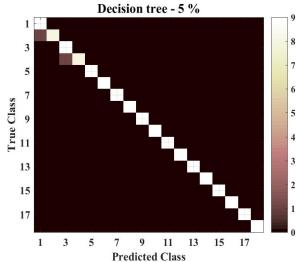
# RIGA TECHNICAL UNIVERSITY Confusion matrix

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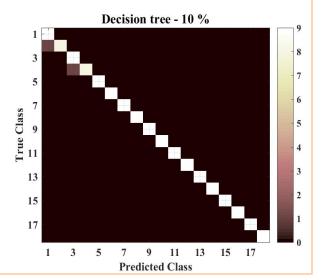


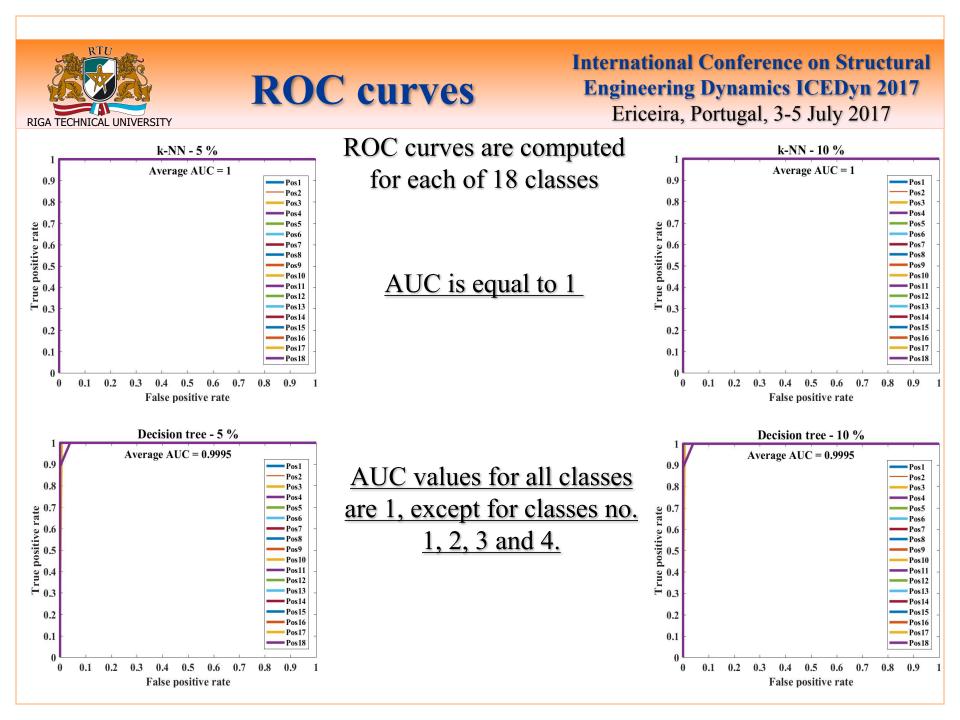
A perfect classification for both damage severities





A slight misclassification in classes no. 2 and 4







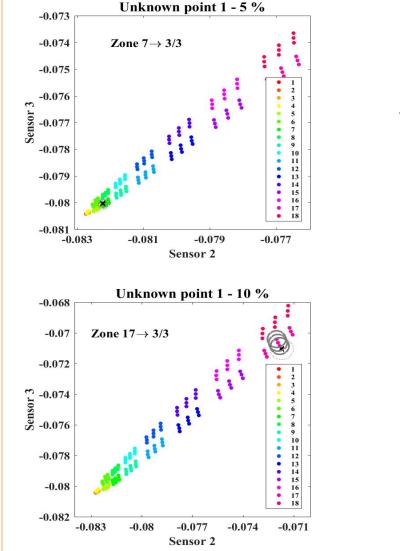
# 2 new points subjected to classification with k-NN and decision trees

Damage severity 10 %			Damage severity 5 %				
$X_1$	0.34	Y <sub>1</sub>	0.005	$\mathbf{X}_{1}$	0.13	$Y_1$	0.035
$X_2$	0.2	$Y_2$	0.05	$X_2$	0.32	$Y_2$	0.07

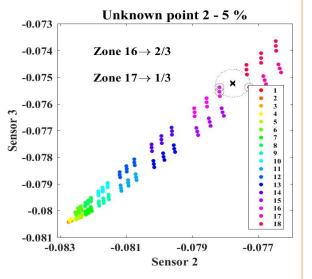
Point 1 – zone 17 Point 1 – zone 7

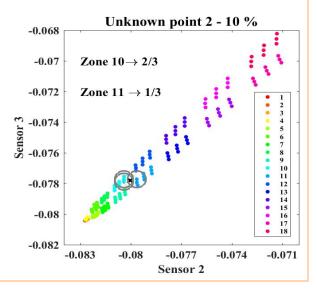
Point 2 – between zones 9, 10, 11 and 12 Point 2 – between zones 16 and 18

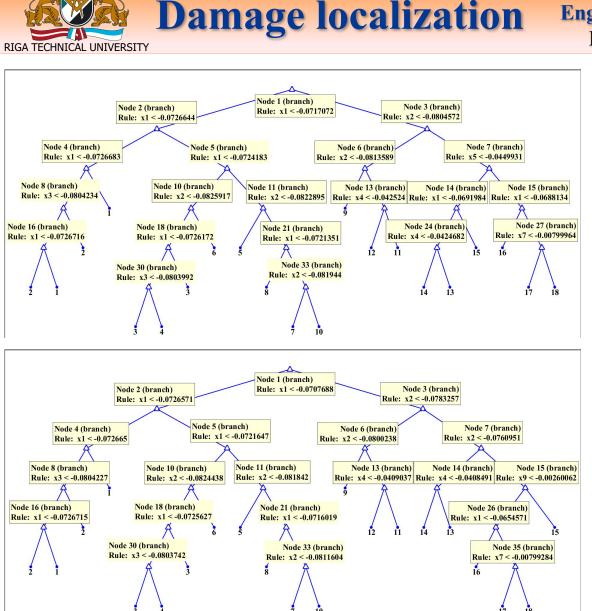




k-NN search









Unknown point 1 – Zone 7 Unknown point 2 – Zone 18

10 %

Unknown point 1 – Zone 17 Unknown point 2 – Zone 9





- The damage localization methodology for plate structures based on data classification with *k*-NN and decision trees is proposed.
- Classification parameters are optimized to minimize the resubstitution and cross-validation errors.
- The performance of classifiers is assessed through ROC curves with accompanying AUC metric and confusion matrices. These metrics suggest a high quality of classification.
- It is found that there is a good agreement between the localization results of both classifiers and these results are in accordance with the actual coordinates of query points for both severities of damage (5 % and 10 %).



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## **Thank You for your attention!**







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