#### **Thesis Presentation**

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#### Content

#### **I** Introduction

- Classic lamination theory(CLT)
- Failure theory
- Evolutionary artificial neural network

#### II New GA Model

- optimal design for cross ply laminate
- optimal design for angle ply laminate

#### III Approximation of CLT and failure theories.

- Data preparation
- Learning
- Experiment

# 1): Classic Lamination Theory

$$\begin{bmatrix} N_{x} \\ N_{y} \\ N_{xy} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} & A_{16} \\ A_{12} & A_{22} & A_{26} \\ A_{16} & A_{26} & A_{66} \end{bmatrix} \begin{bmatrix} \varepsilon_{x}^{0} \\ \varepsilon_{y}^{0} \\ \gamma_{xy}^{0} \end{bmatrix}$$

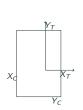
$$+ \begin{bmatrix} B_{11} & B_{12} & B_{16} \\ B_{11} & B_{12} & B_{16} \\ B_{16} & B_{26} & B_{66} \end{bmatrix} \begin{bmatrix} k_{x} \\ k_{y} \\ k_{xy} \end{bmatrix}$$

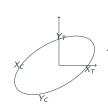
$$= \begin{bmatrix} B_{11} & B_{12} & B_{16} \\ B_{12} & B_{22} & B_{26} \\ B_{16} & B_{26} & B_{66} \end{bmatrix} \begin{bmatrix} \varepsilon_{x}^{0} \\ \varepsilon_{y}^{0} \\ \gamma_{xy}^{0} \end{bmatrix}$$

$$+ \begin{bmatrix} D_{11} & D_{12} & D_{16} \\ D_{11} & D_{12} & D_{16} \end{bmatrix} \begin{bmatrix} k_{x} \\ k_{y} \\ k_{y} \end{bmatrix}$$

 $+\begin{bmatrix} D_{11} & D_{12} & D_{16} \\ D_{11} & D_{12} & D_{16} \\ D_{16} & D_{26} & D_{66} \end{bmatrix} \begin{bmatrix} k_x \\ k_y \\ k_{xy} \end{bmatrix}$ 

## **Introduction: Failure Theory**





Maximum stress failure

$$SF_{MS}^{k} = \min \text{ of } \begin{cases} SF_{X}^{k} = \begin{cases} \frac{X_{t}}{\sigma_{11}}, \text{ if } \sigma_{11} > 0\\ \frac{X_{c}}{\sigma_{11}}, \text{ if } \sigma_{11} < 0 \end{cases} \\ SF_{Y}^{k} = \begin{cases} \frac{Y_{t}}{\sigma_{22}}, \text{ if } \sigma_{22} > 0\\ \frac{Y_{c}}{\sigma_{22}}, \text{ if } \sigma_{22} < 0 \end{cases} \\ SF_{S}^{k} = \begin{cases} \frac{S}{|\tau_{12}|} \end{cases}$$

- **Figure 2:** Schematic failure surfaces for maximum stress and quadratic failure criteria
- Tsai-wu failure theory

$$\begin{split} H_1\sigma_1 + H_2\sigma_2 + H_6\tau_{12} + H_{11}\sigma_1^2 + H_{22}\sigma_2^2 \\ + H_{66}\tau_{12}^2 + 2H_{12}\sigma_1\sigma_2 < 1 \end{split}$$

#### Introduction: Neural network structure

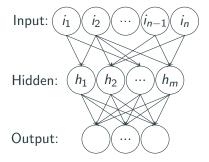


Figure 3: Neural Network Model

#### 2. Traditional GA Model

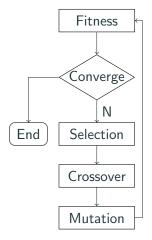


Figure 4: GA process

#### 3. New GA Model

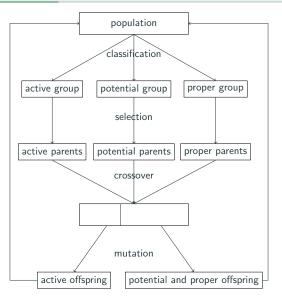
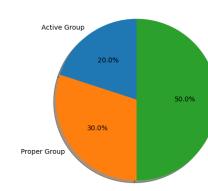


Figure 5: GA Model

#### 3. New GA Parents

- acitve group: individual is used to increase the diversity of the population
- potential group: individual doesn't fulfill constraint
- proper group: individual meet constraint

Figure 6: Parents



### **Experiment for cross ply laminate**

Figure 7: Model for cross ply laminate

0
90
90
0
90

#### **Experiment for cross ply laminate**

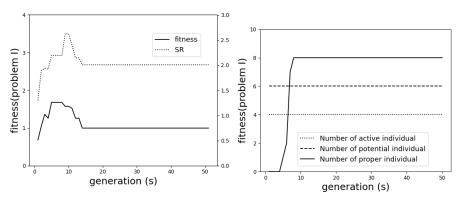


Figure 8: Parents

Figure 9: Parents

#### **Experiment for Comparison**

**Table 1:** The optimum lay-ups for the loading  $N_x=1e6\ N$ 

Cross Ply $[0_M/90_N]$	Choudhury	and Mondal's	Current Research		
Material	Glass-Epoxy	Graphite-Epoxy	Glass-Epoxy	Graphite-Epoxy	
М	68	17	78	18	
N	72	18	28	8	
no. of lamina(n)	140	35	106	26	
SR	2.01	2.10	2.03	2.16	
weight	9.10	1.84	6.89	102.5	

#### Experiment for angle ply laminate: Self-adaptative GA

- Modifying selection strategy: in order to handle the constraint search
- Self-adaptative mutation direction of fiber orientation and laminate thickness: random change the length, and the angle in the laminate.
- The self-adaptative parameters don't refer to parent's proportion, mutation probability.

#### **Experiment for angle ply laminate: mutation operator**

$$md = [CT_1, \cdots, CT_{n-1}, CT_n] - [ICV_0, \cdots, ICV_{n-1}, ICV_n]$$

- md means mutation direction.
- CT<sub>i</sub> denotes the i-th constraint, such as weight, safety factor.
- ICV<sub>i</sub> denotes individual's i-th constraint value, such as, weight, safety factor of current individual.

## **Experiment for angle ply laminate: mutation operator**

• length mutation =

$$\begin{cases} LMC * [0, \sum_{i=1}^{N} md_i] & \text{if } \sum_{i=1}^{N} md_i > 0 \\ LMC * [\sum_{i=1}^{N} md_i, 0] & \text{if } \sum_{i=1}^{N} md_i < 0 \end{cases}$$

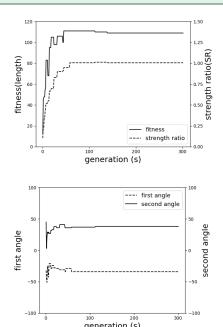
LMC stands for length mutation coefficient, it's a positive integer.

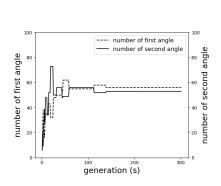
• angle mutation =

$$\begin{cases} AMC*[0,\sum_{i=1}^{N}md_i] & \text{if } \sum_{i=1}^{N}md_i>0\\ AMC*[\sum_{i=1}^{N}md_i,0] & \text{if } \sum_{i=1}^{N}md_i<0 \end{cases}$$

AMC stands for angle mutation coefficient, it's sign is unclear.

# **Experiment:** $N_x = 10, N_y = 5$ MPa m





**Figure 10:** Two distinct angles in the laminate

### Comparison

Table 2: Comparison with the results of DSA

Loading	Akbulut and Sonmez's Study				Present Study				
$N_x/N_y/N_{xy}$ (MPa m)	Optimum lay-up sequences	laminate thickness	TW	MS	Optimum lay-up sequences	laminate thickness	TW	MS	
10/5/0	[37 <sub>27</sub> /-37 <sub>27</sub> ] <sub>s</sub>	108	1.0068	1.0277	[33 <sub>29</sub> /-39 <sub>25</sub> /-39] <sub>s</sub>	109	1.0074	1.0246	
20/5/0	[31 <sub>23</sub> /-31 <sub>23</sub> ] <sub>s</sub>	92	1.0208	1.1985	[33 <sub>22</sub> /-31 <sub>24</sub> ] <sub>s</sub>	92	1.0055	1.2065	
40/5/0	[26 <sub>20</sub> /-26 <sub>20</sub> ] <sub>s</sub>	80	1.0190	1.5381	$[29_{18}/-21_{23}/-\bar{2}1]_s$	83	1.0034	1.7350	
80/5/0	[21 <sub>25</sub> /-19 <sub>28</sub> ] <sub>s</sub>	106	1.0113	1.2213	[-20 <sub>27</sub> /21 <sub>25</sub> /25] <sub>s</sub>	105	1.0029	1.2063	
120/5/0	[17 <sub>35</sub> /-17 <sub>35</sub> ] <sub>s</sub>	140	1.0030	1.0950	[-18 <sub>34</sub> /17 <sub>36</sub> ] <sub>s</sub>	140	1.0000	1.0898	

# Paper 3: Angle ply laminate

$+\theta$
$-\theta$
• • •
$-\theta$
$+\theta$

Figure 11: Model for Angle ply laminate

## Paper 3: General Neural Network

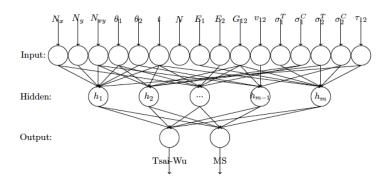


Figure 12: General Neural Network for CLT

# Paper 3: Trainging data preparation

Table 3: Part of train dataset

Input				Output		
Load	Laminate	Material	Failure	MS	Tsai-Wu	
	Structure	Property	Property	IVIS	I SdI-VVU	
-70,-10,-40,	90,-90,4,1.27,	38.6,8.27,0.26,4.14,	1062.0,610.0,31,118,72,	0.0102,	0.0086	
-10,10,0,	-86,86,80,1.27,	181.0,10.3,0.28,7.17,	1500.0,1500.0,40,246,68,	0.4026,	2.5120	
-70,-50,80,	-38,38,4,1.27,	116.6,7.67,0.27,4.173,	2062.0,1701.0,70,240,105,	0.0080,	0.0325	
-70,80,-40,	90,-90,48,1.27,	38.6,8.27,0.26,4.14,	1062.0,610.0,31,118,72,	0.0218,	0.1028	
-20,-30,0,	-86,86,60,1.27,	181.0,10.3,0.28,7.17,	1500.0,1500.0,40,246,68,	0.6481,	0.9512	
0,-40,0,	74,-74,168,1.27,	181.0,10.3,0.28,7.17,	1500.0,1500.0,40,246,68,	1.3110,	3.9619	

### Paper 3: Prediction

Table 4: Comparsion between practical and simulation

Input				Output				
Load	Laminate	Material	Failure	CLT		ANN		
Structure		Property	Property Property		MS Tsai-Wu		MS Tsai-Wu	
-10,40,20	26,-26,168,1.27	116.6,7.67,0.27,4.17	2062.0,1701.0,70,240,105	0.342	0.476	0.351	0.492	
20,-70,-30	10,-10,196,1.27	181.0,10.3,0.28,7.17	1500.0,1500.0,40,246,68	0.653	0.489	0.612	0.445	
60,-20,0	82 -82,128,1.27	181.0,10.3,0.28,7.17	1500.0,1500.0,40,246,68	1.663	0.112	1.673	0.189	

## Paper 3: Prediction

- 1) paper1: journal of reinforced plastic and composites
- 2) paper2: journal of thermoplastic composite
- 3) paper3: which journal?