



Ph.D. Dissertation Defense

Stress-strain Response of High-performance Crystalline Aromatic-aliphatic Polyamides

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Introduction >> High-performance Polymers



Film



www.familyhandyman.com

Frame



www.microsoft.com



- www.tirerack.com
- Excellent mechanical strength per unit weight,
- High thermal stability (continuous use temperatures above 150 °C, combined with low flammability),
- **Chemical** resistance (also at elevated temperatures)
- **Electrical** properties (e.g., as insulators)



Aromatic polyamide





Aromatic polyimide



Kapton



U-polymer

Aromatic heterocyclic polymer



PBT

Polymer Journal 48.2 (2016): 117-138.

Upilex R

Aromatic polysulfone CH₃

PST

Aromatic polytriphenylamine



Introduction >> Aromatic Polyamides

- Aromatic polyamides (aramids), such as poly(p-phenylene terephthalamide) (PPTA), which is also trademarked as Kevlar[®] and Twaron[®]
- Excellent thermal and oxidative stability, high mechanical strength, low flammability and good chemical and radiation resistance.





www.il.ngb.army.mil



www.superleggera.com.au



en.wiktionary.org

• Tensile strength of PPTA is much higher than other commonly seen polymers and even 5 times higher than steel



Introduction >> **PPTA Limitations**

- Melt processing and solution processing
- Hard to process (injection, extrusion, and 3D printing)



Taking up



Introduction >> **Solution**

Reasons for high melting temperature:

- rigid structure of aromatic moieties
- low flexibility of the chains (no rotation)
- strong intermolecular interactions



Solution:

- Increase flexibility of the structure
 - Introducing flexible aliphatic compound to be aromatic-aliphatic polyamide
 - PAPX, X is depending on how many methylene units are included



Introduction >> Experimental Studies

Researchers synthesized polyamides containing various aromatic and aliphatic compounds.[1,2,3] It was found that the melting temperature increases with increasing aromaticity but decreases with increasing length of aliphatic chain. [1,2,3]



However, it was also found that the introduction of flexible aliphatic compounds will cause a reduction in the thermal and mechanical properties (tensile strength, ultimate strain, and Young's modulus). [4,5]



Introduction >> Challenges of Experimental Studies

1. Difficult, time-consuming, and requiring sensitive equipment to experimentally capture minor changes in the molecular structure.

2. Experimentally difficult to apply tensile stress on crystal polymers in the directions that transverse to the chain direction

3. Challenging to obtain large well aligned single crystal and the morphology of the polymers is complicated (e.g., non-crystalline regions and defects)

4. Hard to distinguish the relative contributions of intramolecular (e.g., covalent bonds) interactions and intermolecular interactions (e.g., H-bonding and pi-stacking) to bulk mechanical properties.

Textile Research Journal 87.8 (2017): 984-1010. Polymer Engineering & Science 34.2 (1994): 141-152. US Army Research Laboratory Aberdeen Proving Ground United States, 2018. Polymer 22.7 (1981): 960-965. Journal of materials science 9.11 (1974): 1809-1814. Journal of Materials Science 35.3 (2000): 573-581.





Introduction >> Computational Studies

Time scale vs. Length scale of computational methods



MD simulations have been successfully applied to study aromatic-aliphatic polyamide crystals, including the role of aliphatic and aromatic groups on intermolecular interactions, free volume, and glass transition temperature [1, 2], the origin of melting, and the influence of methylene segments on crystal packing and chain conformation [3].

[1] *Polymer* 81 (2015): 50-61.

[2] Dynamic Behavior of Materials, Volume 1. Springer, Cham, 2014. 187-193.

[3] Macromolecules 49.3 (2016): 950-962.

Introduction >> Limitations of Previous Sim. Studies

• No comparison of a comprehensive set of potentials, including multiple reactive and non-reactive models



- No evaluation of the ability to model H-bonding and $\pi\mbox{-stacking patterns}$



• No generalizability assessed by evaluating homologous material systems



- [1] Journal of Materials Engineering and Performance 20 (2011), pp. 1401-1413
 [2] Journal of Materials Science 46.14 (2011), pp. 4787-4802.
- [3] Journal of Materials Engineering and Performance 22.3 (2013), pp. 681695.
- [4] Advances in Materials Science and Engineering 2013 (2013), pp. 115.
- [5] Journal of Materials Engineering and Performance 22.11 (2013), pp. 3269-3287.
- [6] Journal of Materials Science 49.24 (2014), pp. 82728293.

[7] Polymer 114 (2017): 329-347.

Introduction >> **Solution**

- Deshmukh et al. studied the conformational and structural changes with increasing methylene segment ۲ length in aromatic–aliphatic polyamides
- They found that the melting temperature decreases with increasing methylene length ۲





Methods >> Models of Polyamides

- PAP5 to PAP8 are four example aromatic-aliphatic polyamides
- They all have aromatic and amide groups
- The difference among them is the number of carbons between the aromatic rings



Methods >> XRD Patterns

- The initial structures are manually tuned to match the experimental XRD patterns
- Note that they are not identical because the MD models are perfect crystals





Methods >> Simulation Protocol











Results >> Potential Selection – Lattice

	Unit Cell Lattice Parameters					
Polyamide	a (nm)	b (nm)	c (nm)	α (°)	β (°)	γ (°)
ΡΡΤΑ	0.787	0.518	1.29	90	90	90
PAP5	0.850	0.470	2.48	90	85	90



Macromolecules 49 (2016): 950-962.

- Comparison of the accuracy (error) and stability (deviation) of **7** reactive potentials and **2** non-reactive potentials.
- CVFF has the largest error and especially large error in γ of PAP5

Reactive ReaxFF Potentials:

Mattsson et al. Phys. Rev. B 2010, 81, 054103. Zhang et al. J. Phys. Chem. B 2009, 113, 31, 10770-10778. Kamat et al. J. Phys. Chem. A 2010, 114, 48, 12561-12572. Wood et al. J. Phys. Chem. A 2014, 118, 5, 885-895. Vashisth et al. *J. Phys. Chem. A* 2018, 122, 32, 6633-6642. Liu et al. *J. Phys. Chem. A* 2011, 115, 11016-11022. Budzien et al. *J. Phys. Chem. B* 2009, 113, 13142-13151.

Results >> Potential Selection – H-bonding



- Comparison of O-N radial distribution functions calculated for PPTA and PAP5
- Close-to-reference, tall, and narrow peaks are desirable
- For PPTA, the best potentials are Wood, Budzien, Mattsson, Zhang, Vashisth, and Liu
- For PAP5, the best are OPLS, Wood, Zhang, Vashisth, and Liu

Results >> Potential Selection – π-stacking



- Comparison of O-N radial distribution functions calculated for PPTA and PAP5
- Close-to-reference, tall, and narrow peaks are desirable
- For PPTA, the RDF peaks of Liu and OPLS are the closest to the reference value
- For PAP5,Wood and Liu are the most accurate, with relatively narrow peaks

Results >> System Size Selection

- Identify the smallest model that can be used (to maximize computational efficiency) without simulation artifact
- OPLS doesn't break at all
- Vashisth shows fluctuations between 7-12% strain
- Size 4x4x4 and the ReaxFF Liu potential are suitable for the stress-strain response of PPTA and PAP5



Results >> XRD Patterns after Equilibration



- Comparing the XRD patterns of after equilibration using different force fields with the XRD patterns from the initial structures of PPTA and PAP5
- Obviously, Liu has the retains the initial structures better for both PPTA and PAP5, i.e., reproduce the experimental structures of PPTA and PAP5 better than the other force fields

Results >> Stress-strain Simulation



Results >> Stress & Modulus

- With the selected ReaxFF Liu potential, we performed stress-strain simulations for PPTA and 4 PPTA-related aromatic-aliphatic polyamides (PAP5 and PAP6 are from the Deshmukh paper [1])
- Low-strain modulus lower than high-strain modulus
- Low-strain modulus decreases with increasing number of non-aromatic C atoms
- High-strain modulus is essentially independent of the number of non-aromatic C atoms



^[1] Macromolecules 49.3 (2016): 950-962.

Results >> Explain Modulus Trend

• After equilibration, the chains become wavy, especially for polymers with more non-aromatic C



After Equilibration

PAP8

Initial Structure

After Equilibration

Results >> Explain Modulus Trend

- Side view (yz-plane) of the polymers (only backbone C and N are shown)
- As the number of non-aromatic C increases, the atom cloud spreads out more (more wavy chains)



Results >> Explain Modulus Trend

- The decrease in modulus with chain length is due to increasing waviness of the polymers
- Waviness is due to the methylene groups acting as spacers between the hydrogen-bonded amide groups, which increases the conformational freedom of the polymer chains
- More waviness decreases stiffness because force resistant to stretch has less components in the chain direction



Results >> Explain Transition Trend

- Both single chain and crystals exhibit lower stiffness at low strain than high strain and a gradual increase in stiffness around 5% strain; this indicates the behavior is due to intrachain processes
- For the PAP6 crystal only, there is a sharp transition from low to high strain behavior observed around 10% strain
- This suggests there are odd-even effects and that they are due to interchain processes



Results >> Explain Transition Trend

- The gradual increase in stiffness exhibited by all polymers (except PPTA) correlates well with changes in the dihedral angles, indicating the low strain is accommodated by elongation/rotation of wavy chains
- The sharp transition exhibited by PAP6 and PAP8 correlates with intra chain slip, quantified by ring-ring distance
- These even polymers slip because they have the less stable trapezoidal structure





Results >> Explain Transition Trend

• Movies of strain simulations shown from the y-direction where all atoms except the aromatic rings are faded. Only even polymers (PAP6 and PAP8) exhibit interchain slip.

PPTA







PAP6

PAP7





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Results >> Ultimate Stress



- The ultimate stress exhibits odd-even behavior where odd numbers of carbon atoms have larger strength
- Since the chains are extended at this point, and DFT results showed the bond strength is the same for all polymers, so this must be an interchain effect
- Interchain strength is determined by the coplanarity of the rings, quantified by ring-ring angle, which is lower (more aligned rings) for the polymers with odd numbers of carbon atoms, explaining their strength

Results >> Trapezoid and Parallelogram

- Representative snapshots of trapezoidal and parallelogrammatic structures in the backbones of odd (PAP5) and even (PAP6) polymers.
- It is an important origin of odd-even effect



Summary of Polyamides MD Simulations

400

(ÿ) z

-3 +



Suggestions for Future Work



- PPTA and PAP5-PAP8 can be written in the same formula with different X₁ and X₂ groups
- X₁ and X₂ groups for PPTA and PAP5-PAP8 are shown in the following lists
- It means that the functional groups X₁ and X₂ groups affect mechanical properties



Suggestions for Future Work

Effect of Functional Groups on Stress-strain Response



Functional Groups X₁:

hPAP-2	hPAP-12	hPAP-14	hPAP-14-f	hPAP-16	hPAP-23
O S S S S S S S	<u>چ</u> S - چ			\$ 0 \$	
hPAP-idx1	hPAP-idx1-f	hPAP-bipheny	l hPAP-Et	hPAP-Pr	hPAP-NH
	O N H				H N S
hPAP-carbonate	hPAP-urethan	e hPAP-ureth	ane-f hPAP-	urea hPAP-X	6 hPAP-X7
			N N H	O CH _a	3 OH

- Introducing more functional groups to study the effects on mechanical properties
- A list of 18 potential functional groups are listed on the left
- X₂ can be methylene groups with different lengths (e.g., 0-17 carbon atoms)
- Data volume: 18 x 18 = 324 different polyamides

Suggestions for Future Work

input

> Using machine learning to build quantitative structure-properties relationship (QSPR)

Polyamide Information

- Functional groups
- Chain length?
- Molecular weight?
- # of atoms?
- Bonds?



Machine Learning

Mechanical Properties



Low-strain modulus

- High-strain modulus
- Ultimate strain
- Ultimate stress

ML Models

- Linear Regression
- Polynomial Regression
- Gaussian Process Regression
- Artificial Neural Network (ANN)
- Support Vector Machines (SVM)
- Deep Learning
- XGBoost

Summary of Contributions

- Established a force field selection framework that can be applied to other polymeric crystals;
- Provided insight into the important fundamental question of how the processibility of high-performance polyamide PPTA can be improved by modification without compromising the mechanical properties;
- Provided new insight into how inter and intrachain interactions affect the stress-strain response at the atomic scale;
- Introduced a quantity, "waviness", to describe the flexibility of polyamides and discovered the correlation of it with low-strain modulus
- Emphasized the critical role of the odd-even effect observed in both the elastic and ultimate properties of polyamides;
- Explained the origin of the odd-even effect by attributing it to the structure of hydrogen bonding network.



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ExonMobil



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- Dr. Wenjun Li





Dr. Manesh Gopinadhan





Summary of Accomplishments (Master's & PhD):

Publications (10, including 7 first-author):

- Journal Papers: published (9), submitted (1)
- > Patent: 1
- Software Copyrights: 3

Conferences (6):

Presentation (4), posters (2)

Awards:

- Faculty Mentor Program Fellowship (2022)
- UC Merced Bobcat Fellowship (2019, 2020, 2021, 2022)
- Third Place Poster Award of The Web Seminar Series on Tribology (WeSST) (2020)
- MACES Summer Fellowship (sponsored by NASA) (2019, 2020)
- UC Merced Travel Fellowship (2019, 2022)
- South China University of Technology First-tier Scholarship (2016, 2017)
- South China University of Technology Third-tier Scholarship (2015)

Peer Review Service for Journals (11):

Tribology Letters (4), Sustainability (4), Applied System Innovation (2), Applied Sciences (1) **Conference Session Chairs (2):**

Tribochemistry and Biotribology sessions at the 73rd STLE Annual Meeting & Exhibition in Chicago 2019

Thank you for attention!



Appendix



Motivation >> Similarity of Polyamides



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• Only part of the data is shown here

Results >> Mechanical Properties from MD

350 -

300

450

 Low-strain modulus, high-strain modulus, ultimate strain, and ultimate stress



Results >> **Descriptors & Properties**

ID	Descriptor	ID	Descriptor	
x_0	Equilibrium Waviness (Å)	x_9	# of O in X_2	
x_1	# of Methylenes in X_1	x_10	# of C in X ₂	
x_2	# of Rings in X ₁	x_11	a (nm)	
x_3	# of S in X_1	x_12	b (nm)	
x_4	# of O in X ₁	x_13	c (nm)	
x_5	# of C in X ₁	x_14	α (°)	
x_6	# of Methylenes in X ₂	x_15	β (°)	
x_7	# of Rings in X ₂	x_16	γ (°)	
x_8	# of S in X ₂			
ID	Mechanical Property			
y_0	Ultimate Strain (%)			
y_1	Ultimate Stress (GPa)			
y_2	Low-strain Modulus (GPa)			
y_3	High-strain Modulus (GPa)			

For convenience, the descriptors and mechanical properties are represented as IDs (x and y)

•



Results >> **Descriptors & Properties**

- Polynomial Regression using Scikit-Learn package in Python to predict mechanical properties of polyamides
- Tried different combinations of features (descriptors)
- Tried different degrees (1-3) of polynomial functions
- Data is separated into training set (9) and testing set (4)
- 'x_1', 'x_2', 'x_3', ... 'x_14', 'x_15', 'x_16' represent the descriptors

Example of calculation results:

Descriptor Combination	Testing Set Low-strain Moduli (GPa) (hPAP-idx, PAP5, hPAP-12)	Degree of Polynomial	Root Mean Square Error (RMSE) (GPa)
Actual Value	10.0, 174.0, 8.0	-	-
['x_1']	96.5, 92.3, 96.5	1	85.6
['x_2']	60.0, 60.0, 60.0	1	77.8
•••	•••	•••	•••
['x_1', 'x_5', 'x_6', 'x_7', 'x_8']	10.6, 163.5, 10.6	1	36.7
•••	•••	•••	•••
['x_0', 'x_1', 'x_2', 'x_6', 'x_7', 'x_14', 'x_15']	10.6, 163.5, 10.6	2	1.99 Best !

Results >> Structure-Properties Relationship



Low-strain Modulus	High-strain Modulus	Ultimate Strain	Ultimate Stress	 Ultimate Strain (%) Ultimate Stress (GPa) 	
Most Relevant Descriptors				Low-strain Modulus (GPa) High-strian Modulus (GPa) High-strian Modulus (GPa)	
Equilibrium Waviness (Å)	Equilibrium Waviness (Å)	Equilibrium Waviness (Å)	# of Methylenes in X_1	9 325 -	
# of Methylenes in X_1	# of Methylenes in X_1	# of Methylenes in X_1	# of S in X_1	hPAP-X1	
# of Rings in X_1	# of C in X ₂	# of Rings in X ₁		hPAP-X2 hPAP-X2 hPAP-X1 hPAP-X1	
# of Methylenes in X_2	a (nm)	# of S in X_1		25 - hPAP-X2	
# of Rings in X ₂		# of Methylenes in X_2		0	
α (°)				0 50 300 350 400 Actual Value	
β (°)				 Four data points in each 	
Root Mean Square Error of Best Model				color represent four	
1.99 GPa	5.16 GPa	2.20 %	1.36 GPa	materials in the testing set	
Best Model					
2nd degree polynomial	1st degree polynomial	3rd degree polynomial	2nd degree polynomial		

Next Steps >> More Data, Descriptors & ML Models

More data:

• Expanding the data pool by varying the length of methylene groups (300+ more)

 $X_2 = n$ methylenes

Polyamides from literature/database (300+)

Descriptors:

 Calculate dynamics (3D) descriptors from trajectories of MD simulations by developing a Python Package [1]

ML Models:

- Linear Regression
- Polynomial Regression
- Gaussian Process Regression
- Artificial Neural Network (ANN)
- XGBoost



Mechanical Properties

- Low-strain modulus
- High-strain modulus
- Ultimate strain
- Ultimate stress

[1] **Yang**, Panwar, and Martini. PyL3dMD: Python LAMMPS 3D Molecular Descriptors. (To be submitted)