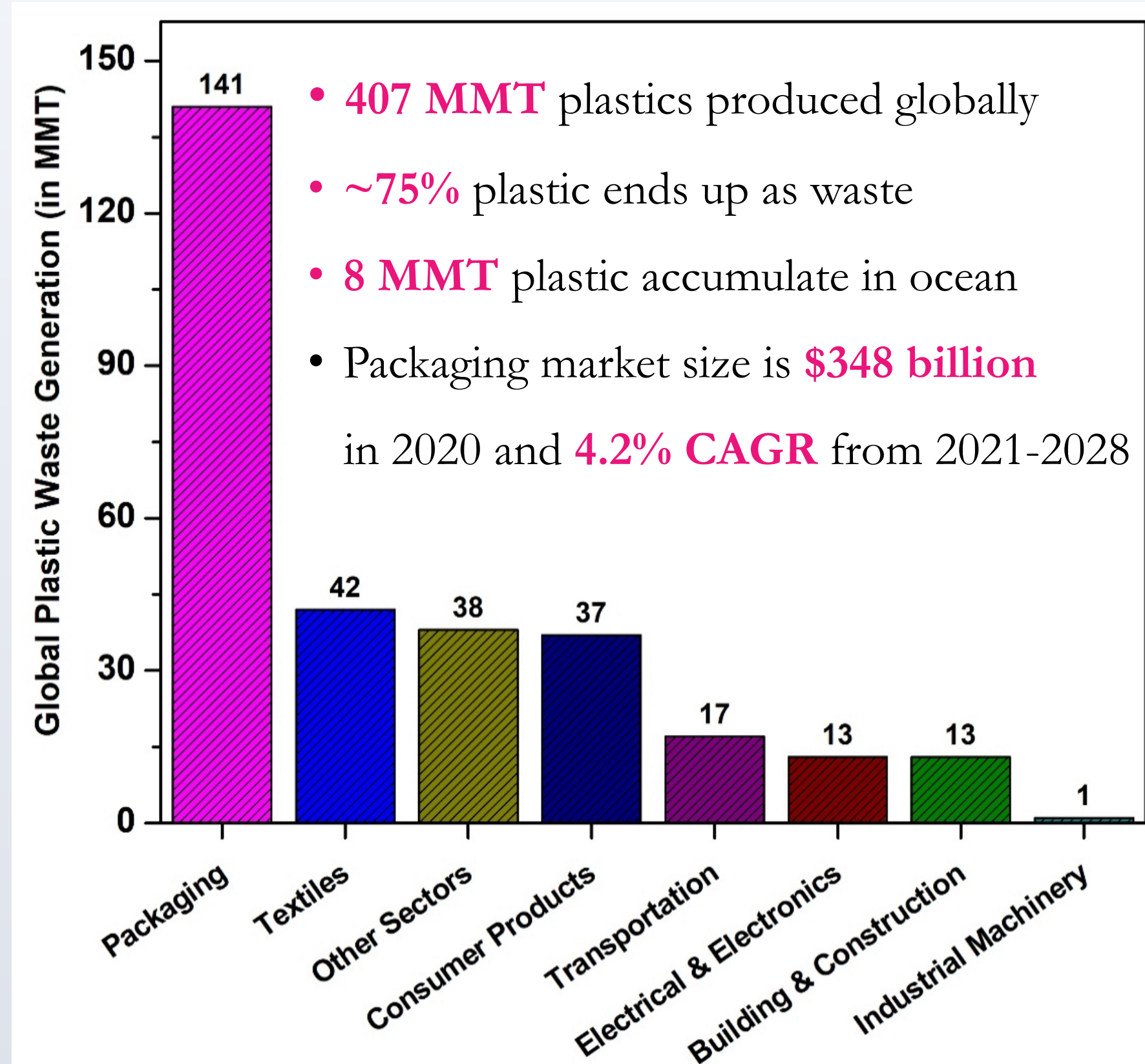
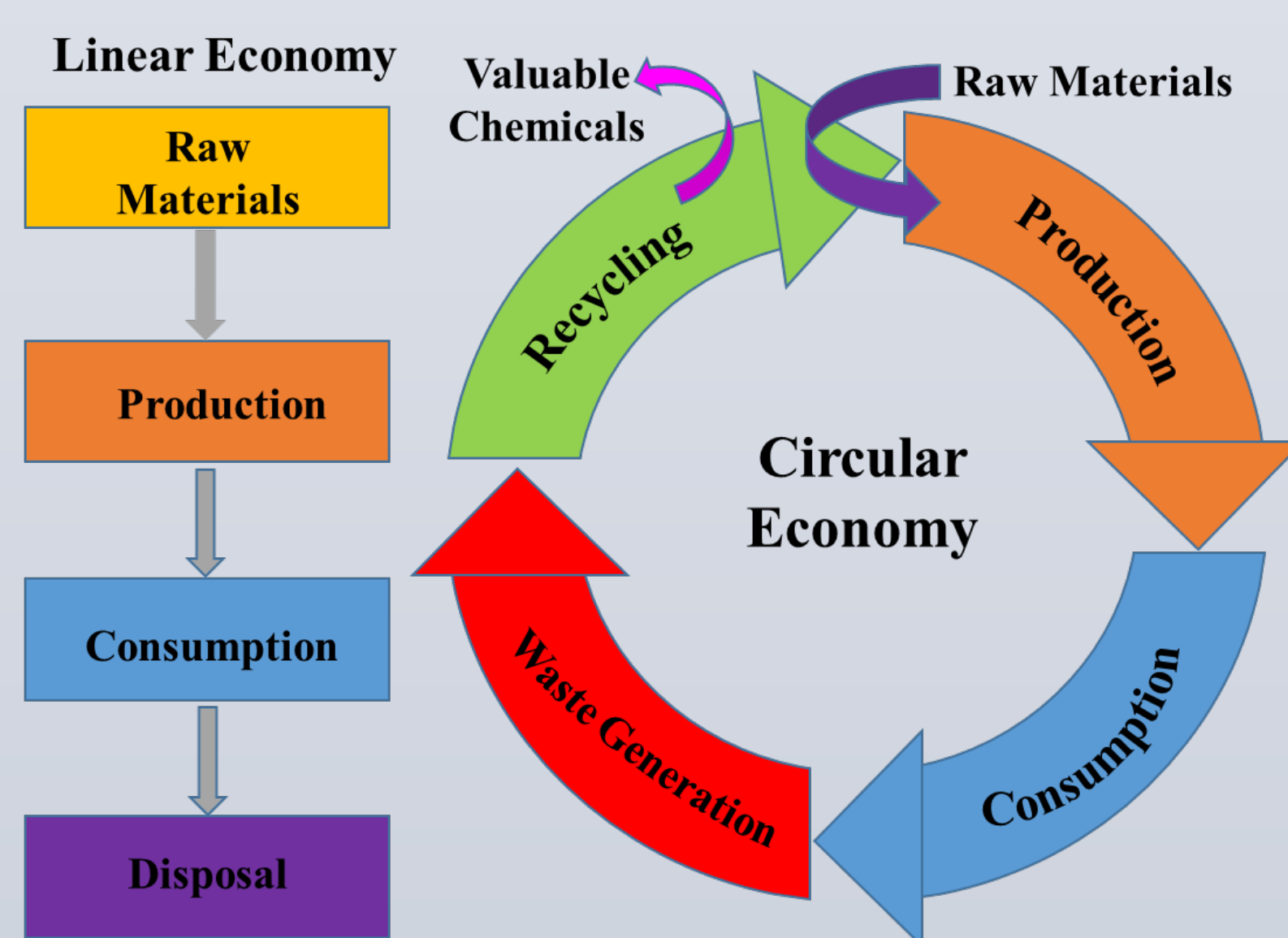


## INTRODUCTION



## Linear versus Circular Economy



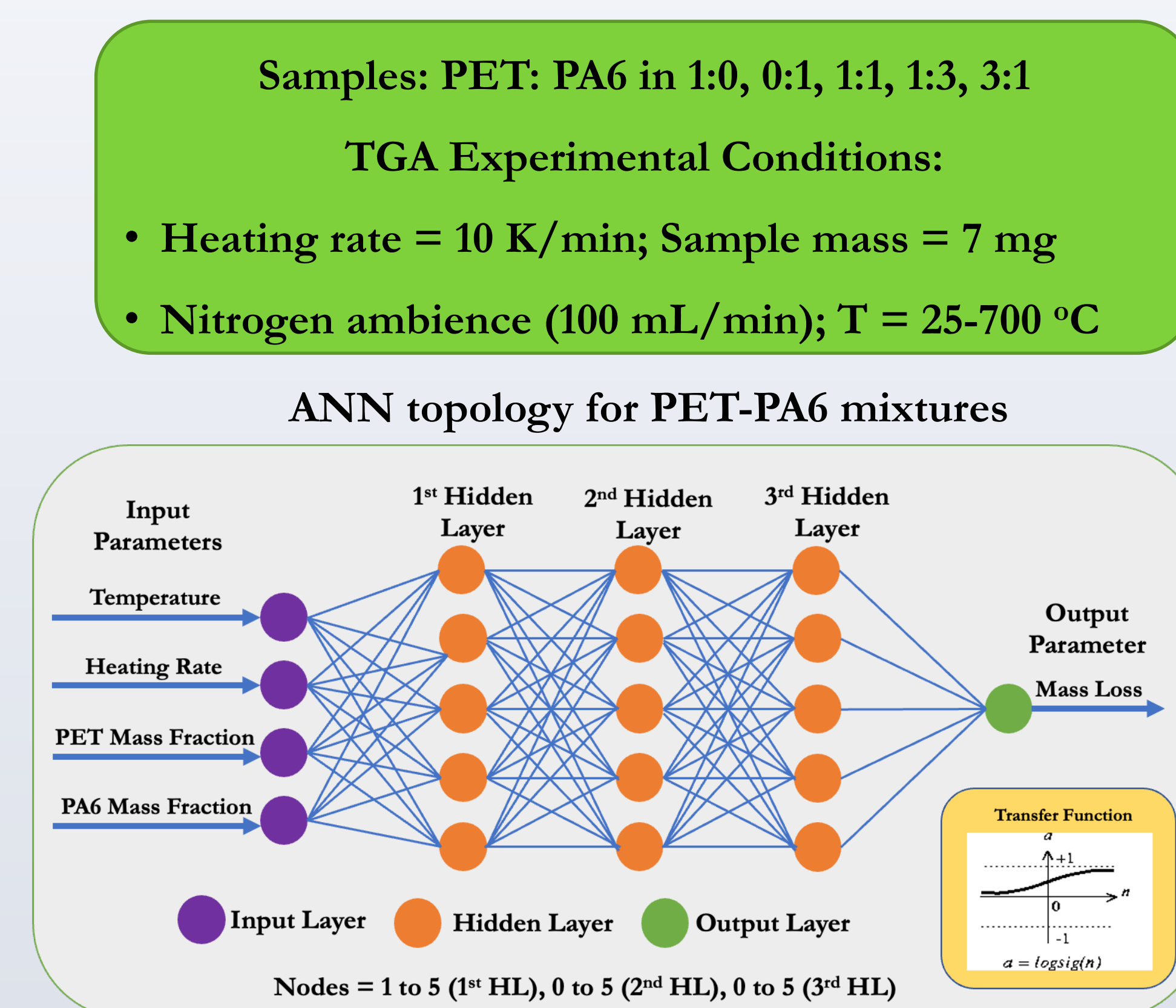
## Plastic Waste Management



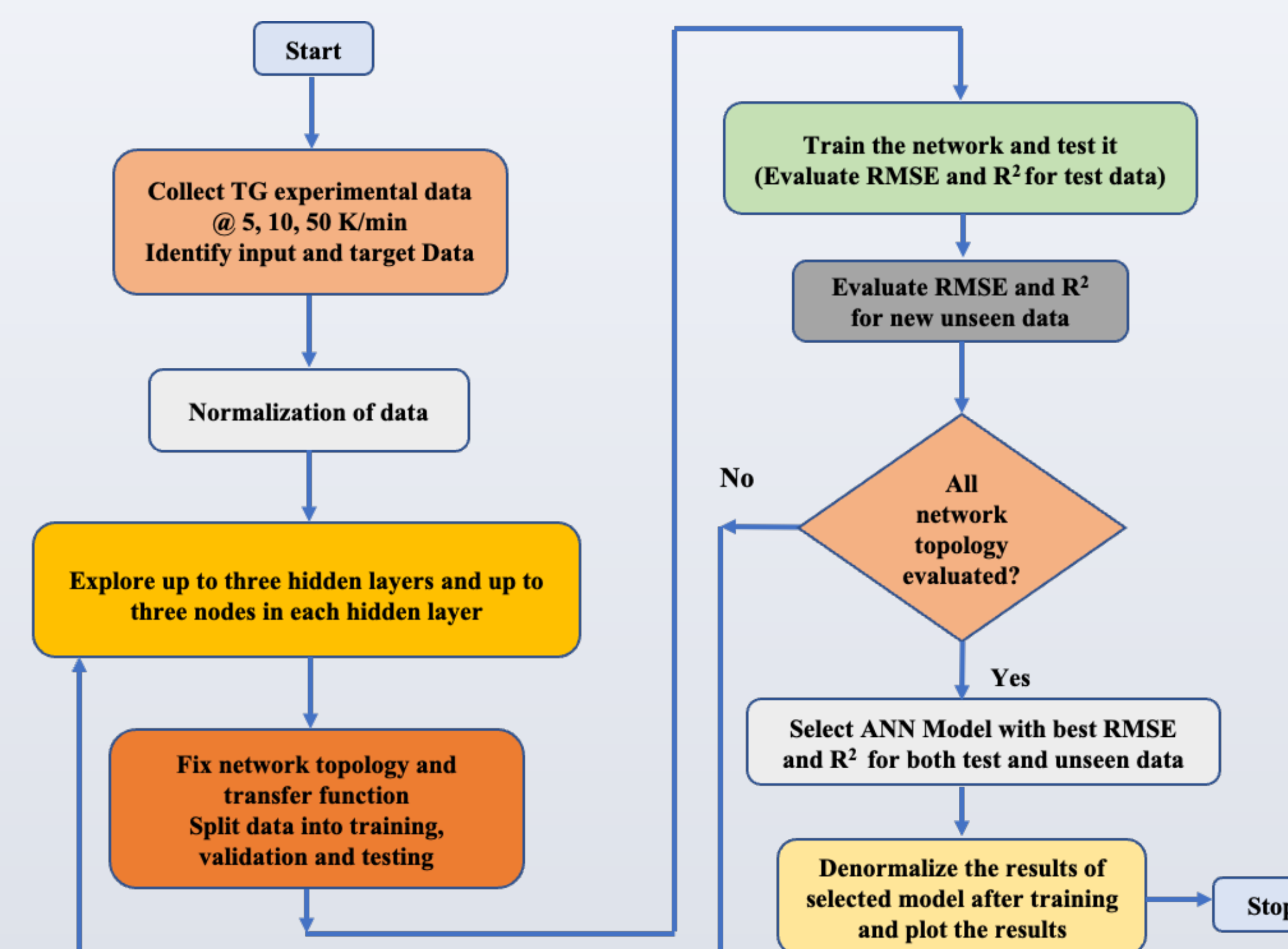
## OBJECTIVES

- To study the pyrolysis behaviour of Polyethylene Terephthalate (PET) and Polyamide 6 (PA6) in different mixture ratios
- To evaluate the kinetic and statistical parameters and model the PET-PA6 mass loss behaviour using artificial neural network (ANN)
- To test the efficacy of ANN model on new dataset

## MATERIAL & METHODS

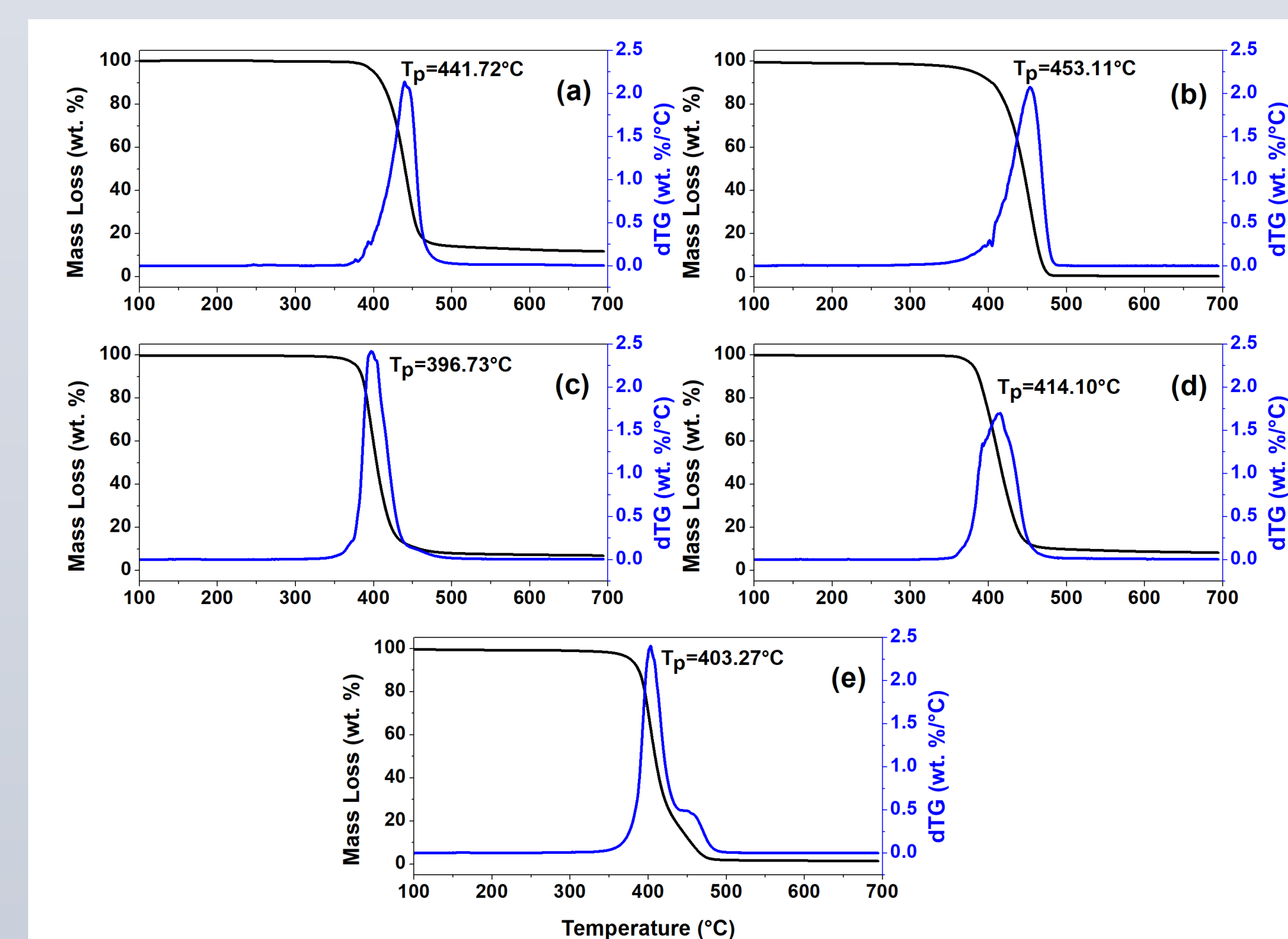


## Flowchart for algorithm of ANN modeling

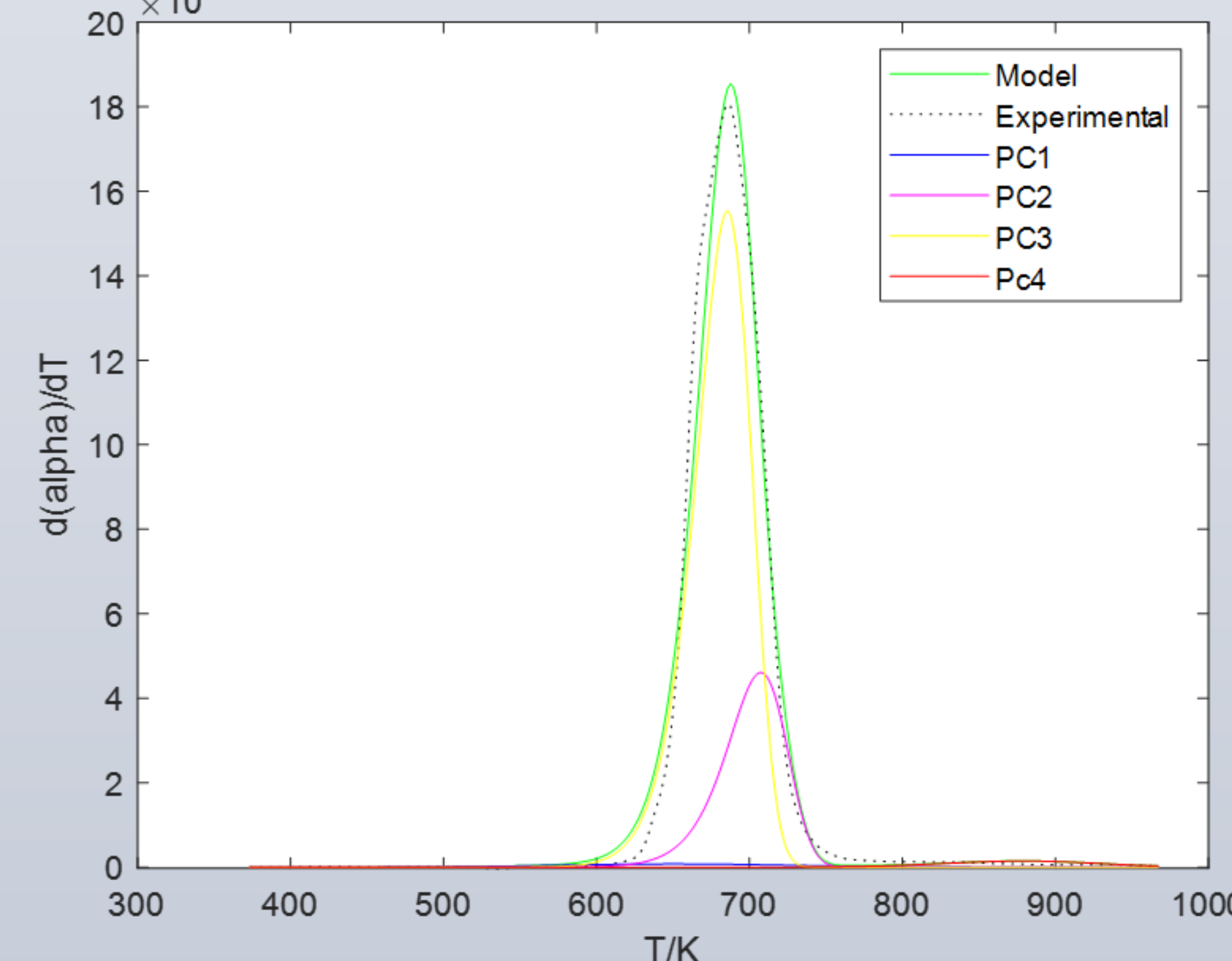


## RESULTS

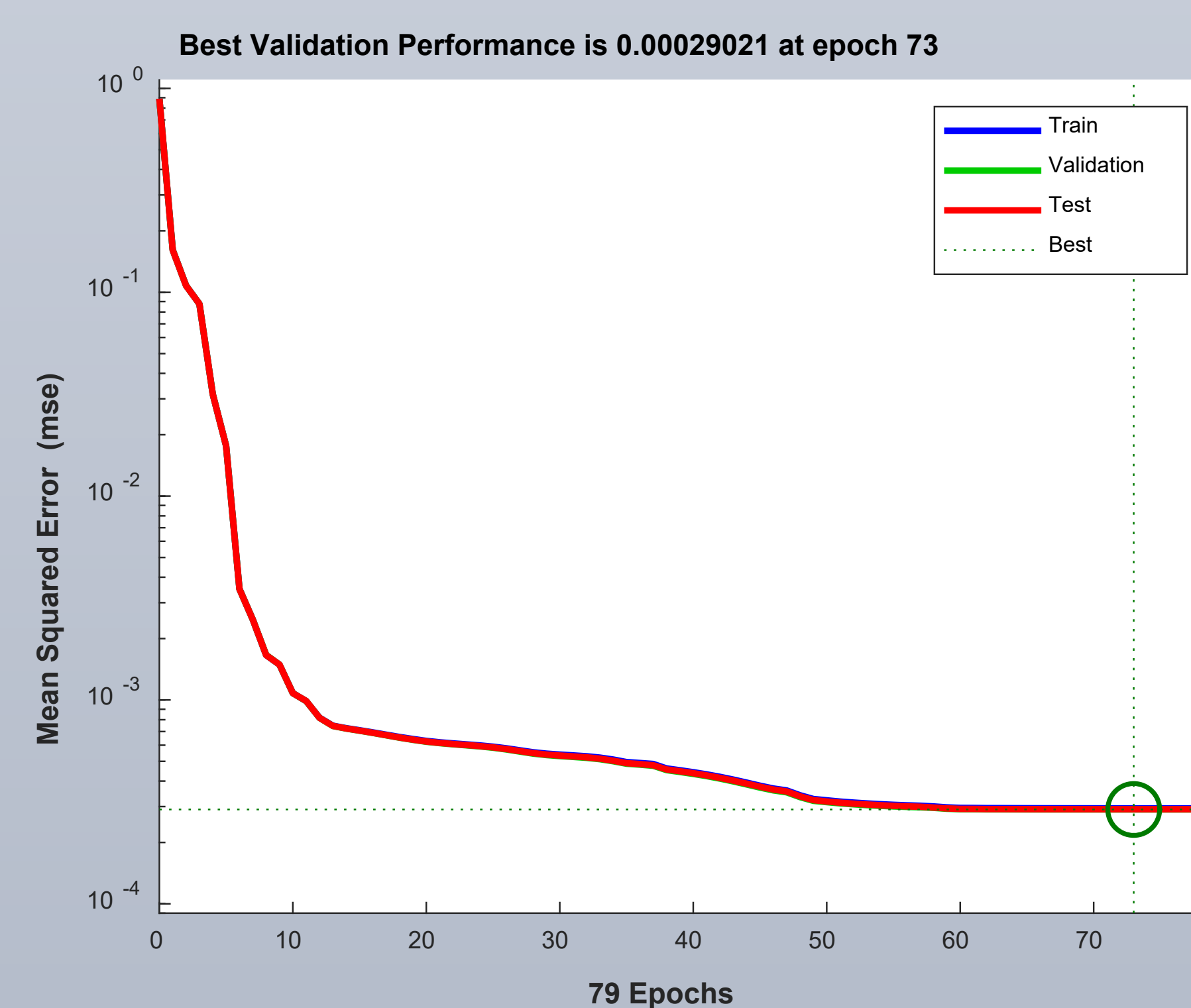
Experimental TG data for PET:PA6 (a) 1:0, (b) 0:1, (c) 1:1, (d) 3:1, (e) 1:3.



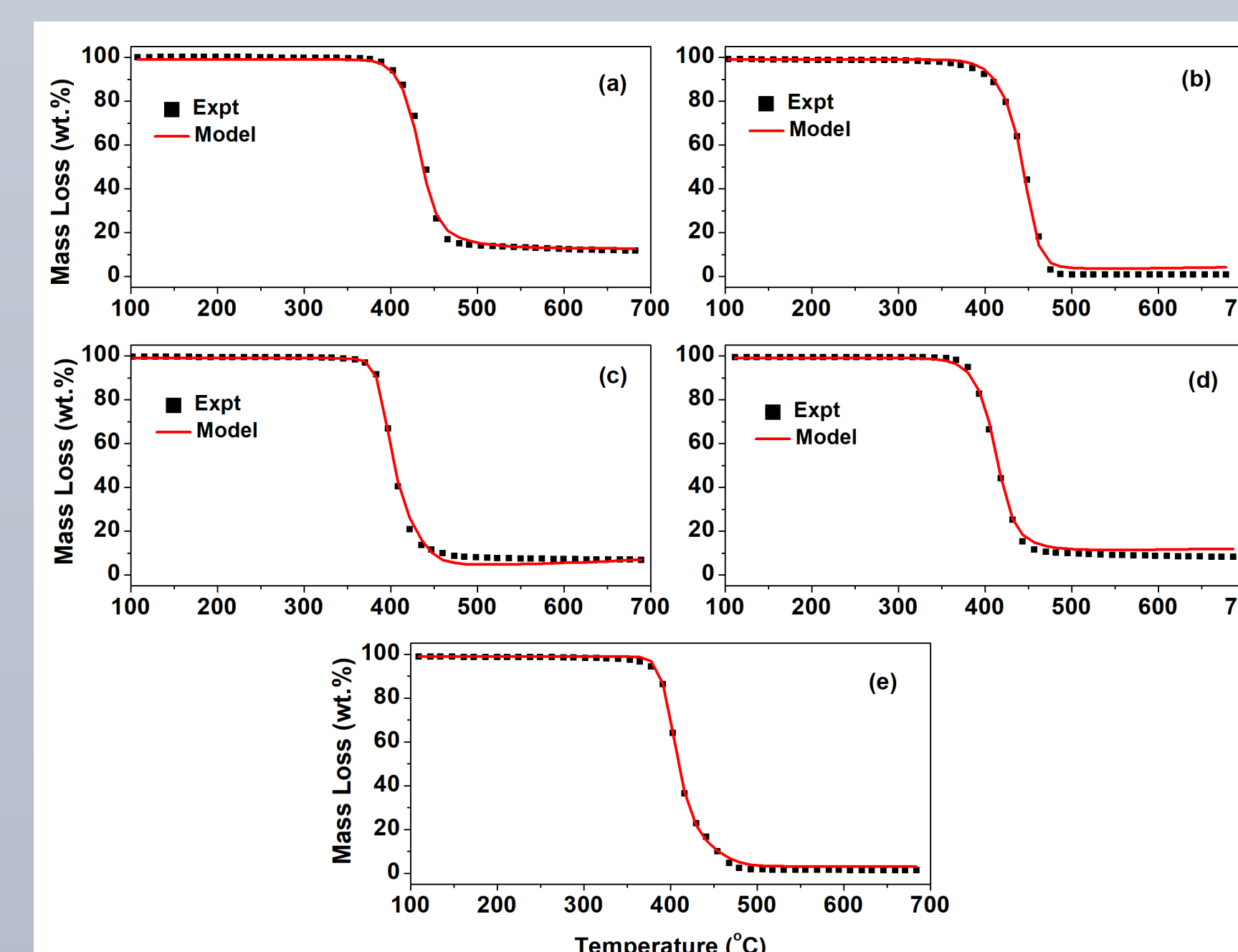
Experimental and simulated DTG profile for PET:PA (3:1) at 10 °C min<sup>-1</sup>.



Performance plot of ANN511 model. The training, validation and testing plots overlap with each other.



Experimental versus predicted TG curves on testing data at 10 °C/min using ANN511 for PET:PA6 mixtures (a) 1:0, (b) 0:1, (c) 1:1, (d) 3:1, and (e) 1:3.



## CONCLUSIONS

- The TGA results confirm that there are **significant interactions** among PET and PA6 mixtures.
- The kinetic study was carried out using isoconversional methods such as Kissinger-Akahira-Sunose (KAS) and Flynn-Wall-Ozawa (FWO), and a **multi-Gaussian distributed activation energy model (DAEM)**.
- The average activation energy ( $E_a$ ) obtained by KAS and FWO methods was in the range of **168-194 kJ mol<sup>-1</sup>**, which is in good agreement with the value obtained by the DAEM model.
- The mass loss profiles of PET-PA6 mixtures were modelled using **Artificial Neural Network (ANN)**.
- The model ANN511 predicted the experimental mass loss behaviour reasonably well for both test data and new data at interpolated conditions of PET:PA6 in different ratios with root mean square error (RMSE) less than 3%.

## REFERENCES

1. Ni, Z., Bi, H., Jiang, C., Wang, C., Tian, J., Zhou, W., & Lin, Q. (2021). Investigation of the co-pyrolysis of coal slime and coffee industry residue based on machine learning methods and TG-FTIR: Synergistic effect, kinetics and thermodynamic. *Fuel*, 305, 121527.
2. Al-Yaari, M., & Dubdub, I. (2020). Application of Artificial Neural Networks to Predict the Catalytic Pyrolysis of HDPE Using Non-Isothermal TGA Data. *Polymers*, 12(8), 1813.
3. Sun, H., Bi, H., Jiang, C., Ni, Z., Tian, J., Zhou, W., ... & Lin, Q. (2022). Experimental study of the co-pyrolysis of sewage sludge and wet waste via TG-FTIR-GC and artificial neural network model: Synergistic effect, pyrolysis kinetics and gas products. *Renewable Energy*, 184, 1-14.
4. Özveren, U., Kartal, F., Sezer, S., & Özdoğan, Z. S. (2022). Investigation of steam gasification in thermogravimetric analysis by means of evolved gas analysis and machine learning. *Energy*, 239, 122232.
5. Wilamowski, B. M., & Yu, H. (2010). Improved computation for Levenberg–Marquardt training. *IEEE transactions on neural networks*, 21(6), 930-937.

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