

Polymer Design using ML

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[Start presentation](#)

CIRCULAR PLASTICS AT TNO



1
SYSTEM INTEGRATION
VALUE CHAIN DESIGN
AND LIFE CYCLE
ASSESSMENT



2
DESIGN
FOR / FROM
RECYCLING



3
PRETREATMENT



4
RECYCLING



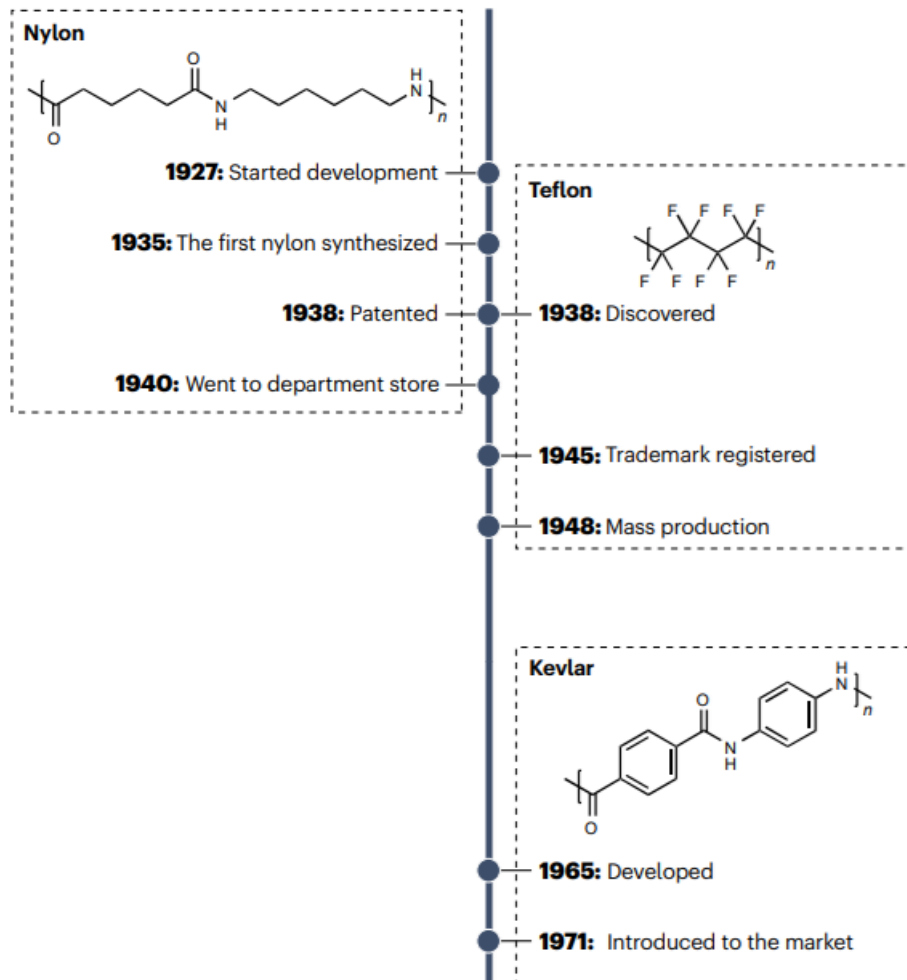
5
MICROPLASTICS

Polymer innovations over the past two centuries

a Selected milestones of polymer science

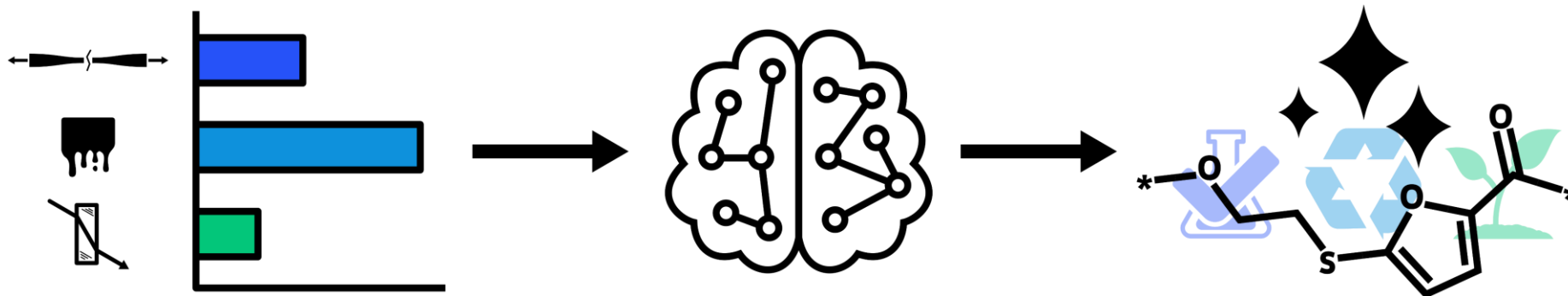
- **1833:** Berzelius coins the term 'polymer'
- **1844:** Goodyear patents the vulcanization of rubber
- **1846:** Schonbein synthesizes nitrocellulose
- **1856:** Parkes patents the first thermoplastic
- **1892:** Cross, Bevan and Beadle patent 'viscose'
- **1907:** Baekeland creates the first synthetic plastic
- **1912:** Klatte invents the production process of PVC
- **1920:** Staudinger publishes 'uber polymerization'
- **1933:** Fawcett and Gibson discover LDPE
- **1937:** Bayer discovers polyurethane
- **1938:** Carothers patents nylon, i.e., Polyamide 6.6, and Plunkett discovers Teflonane
- **1944:** McIntire discovers polystyrene
- **1954:** Ziegler-Natta catalyst discovered for polypropylene production
- **1965:** Kwolek creates Kevlar
- **1977:** Heeger, MacDiarmid and Shirakawa develop polyacetylene, the first conductive polymer
- **2011:** Launch of Materials Genome Initiative
- **2020:** 100-year anniversary of modern polymer science

b Typical deployment progress of some notable polymers

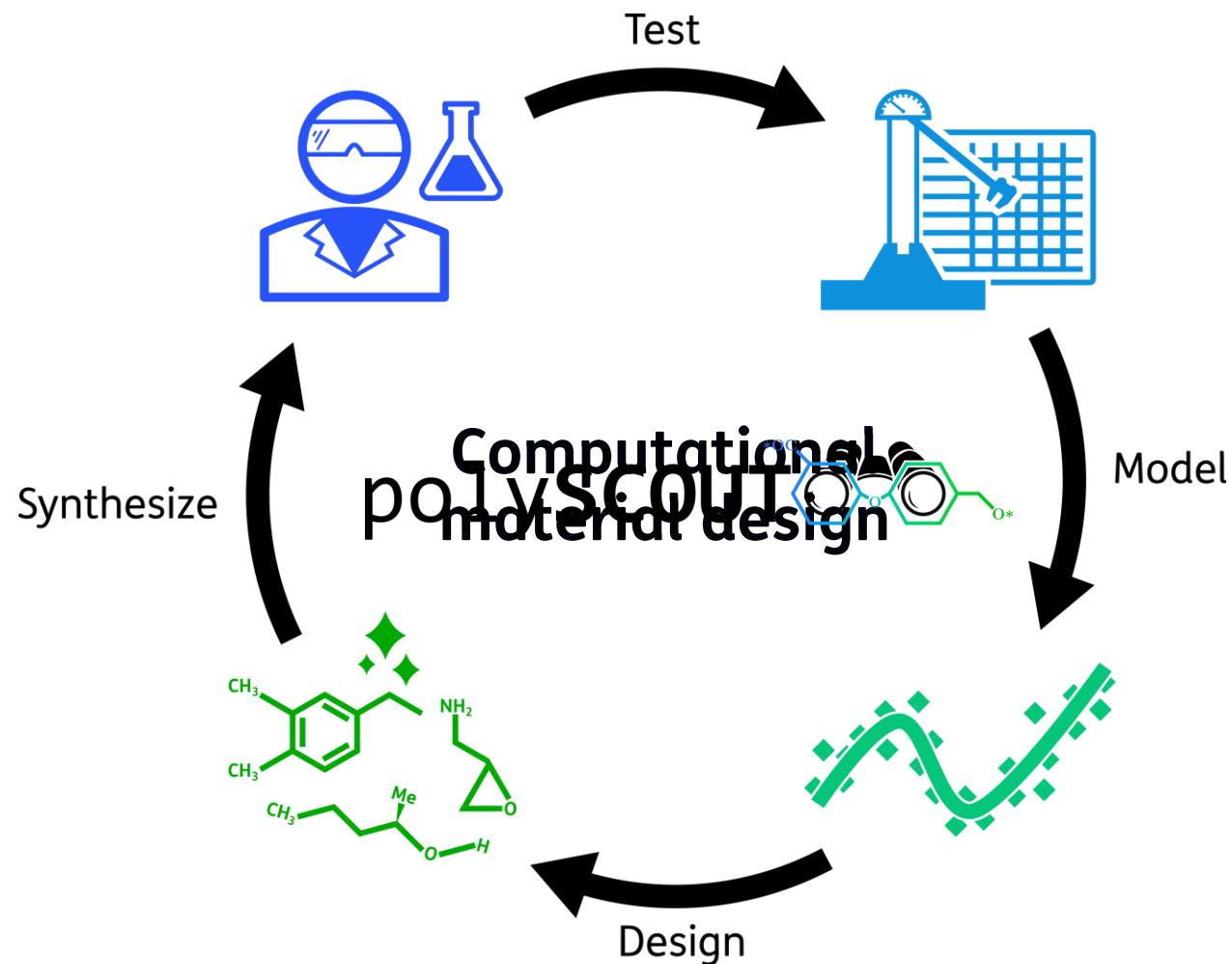


polySCOUT

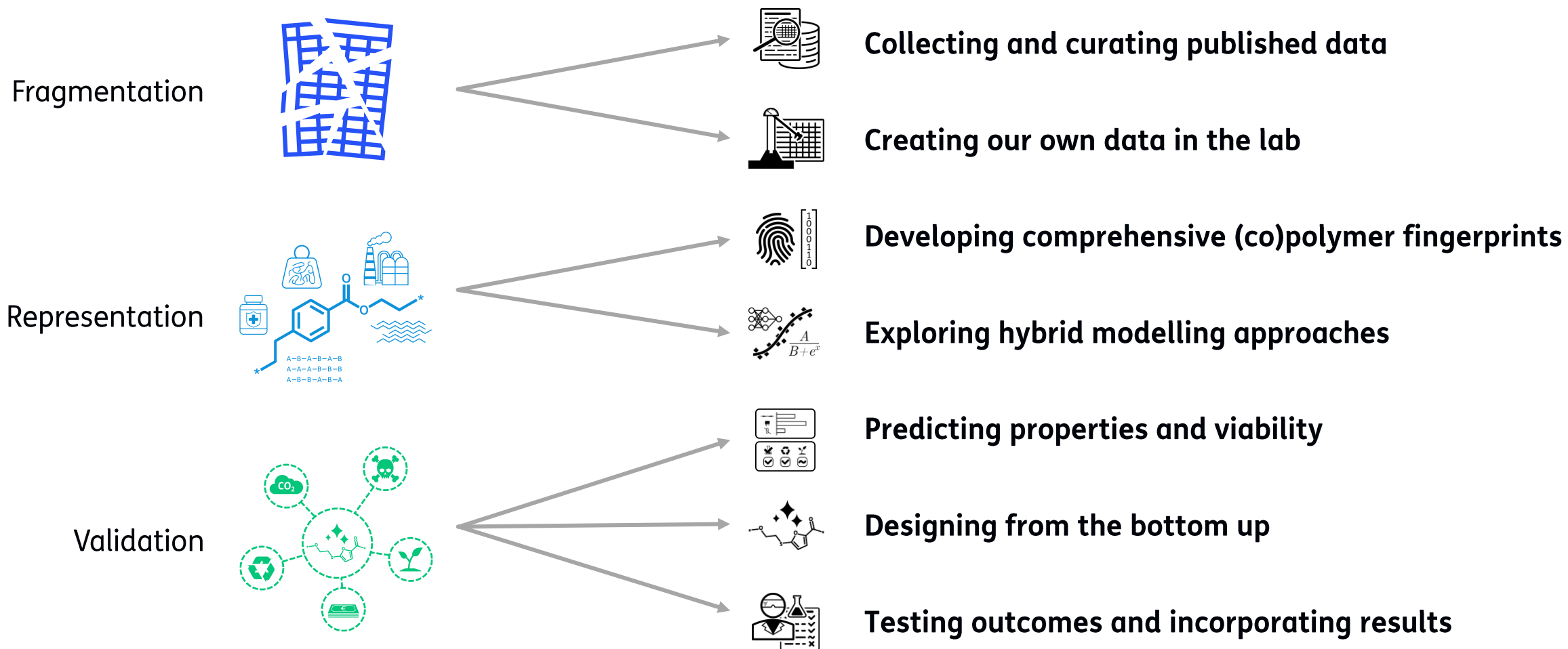
Goal: Develop fast machine learning-based methods that realistically capture the complexity of polymers to design novel candidates matching with requirements for future-proof plastics



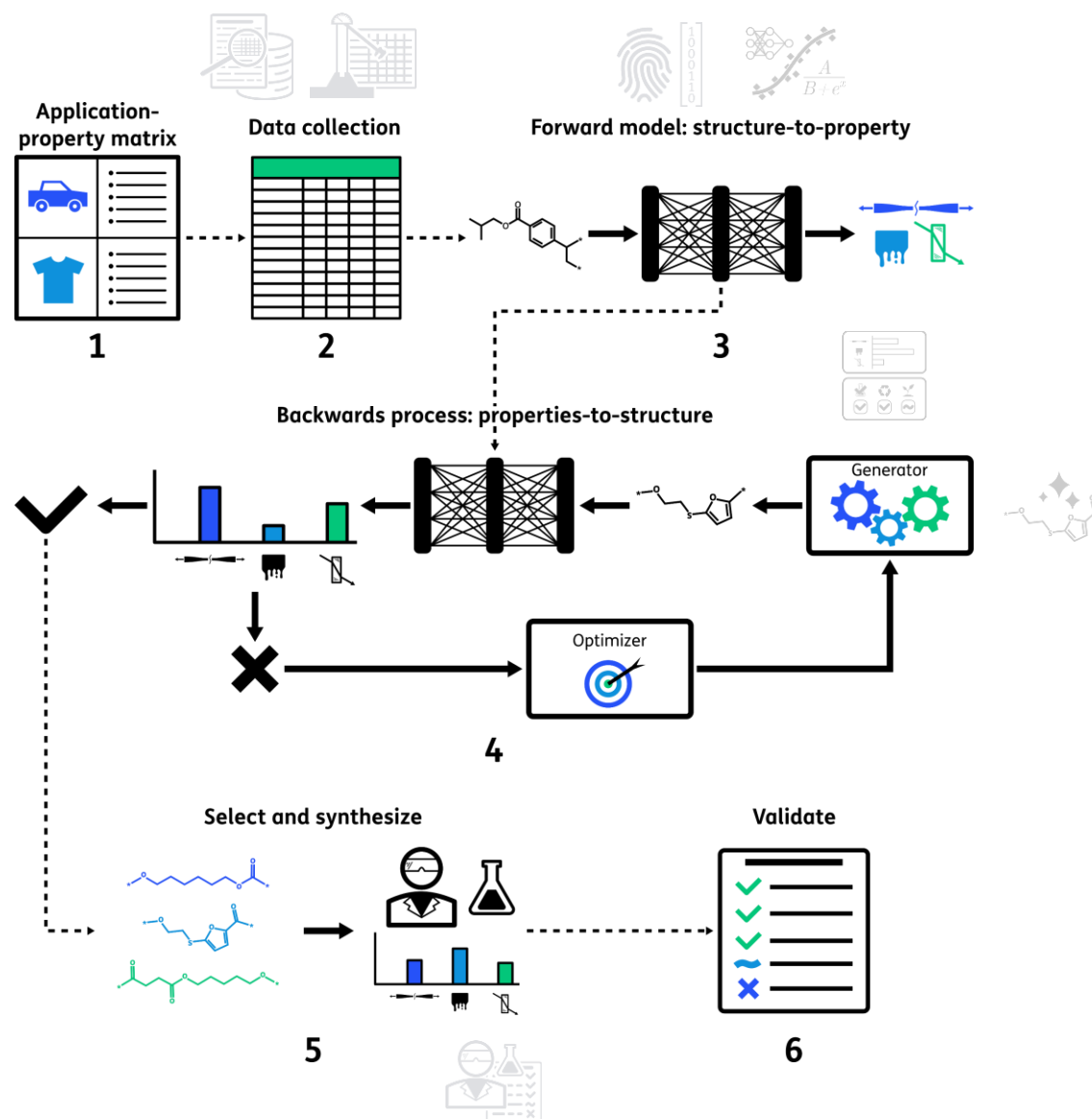
Polymer Informatics @TNO



Addressing the challenges



Our workflow



Application-property matrix



Overview of conventional polymer types and applications

Literature study on existing biopolymers

Textile and packaging identified as first targets

35-40% polyesters



PEBF₆₀SF₄₀ =
Poly(ethylene furanoate-co-sulfanediylidifuranoate)

[O-]C(=O)C1=CC(C)C(=CC(=C1)*)

Match polymer ID to properties

25-35% others

5-10% polyurethanes

↑ Manual validation

Currently for manual validation

Currently properties for ~6000 polymers (and counting) in database

Data creation



High-throughput screening of 8 properties:

- compression modulus
- hardness
- yield strength
- scratch resistance
- storage and loss modulus
- glass transition temperature
- impact strength

To be incorporated in active learning loop with data-driven models

Experimental work: Filling the data base

- Initial materials that will be assessed:

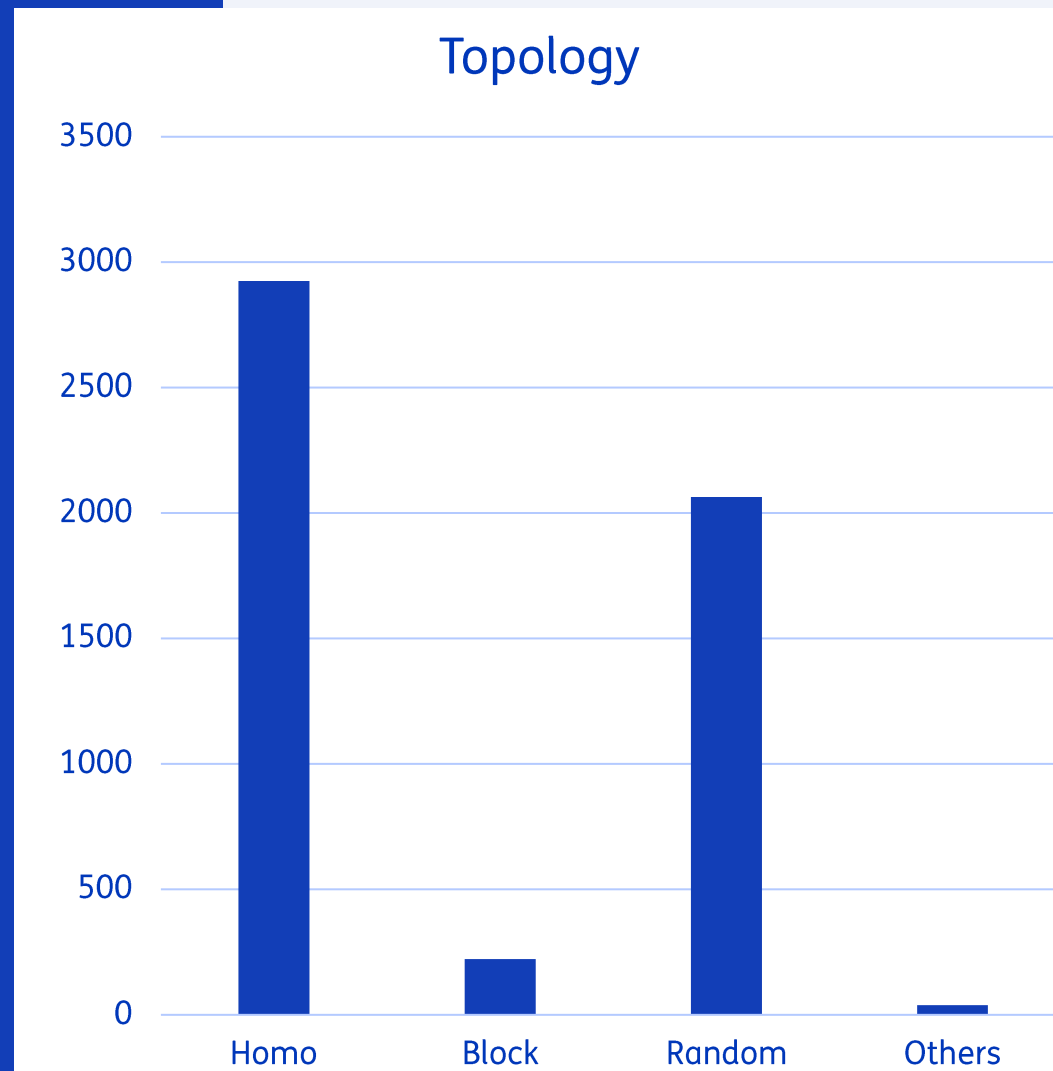
- PET – high IV
- PET – medium IV
- PET – low IV
- PA6.6
- PBS
- PBAT
- PHA
- PLA
- Polytetrahydrofuran
- Polypropyleneglycol
- Polyethyleneoxide
- Polyethyleneglycol
- PA6
- PA11
- PA12
- PA66
- Polyglycolic acid
- Polylactic acid-L
- Polylactic acid-D
- Polylactic acid-L/D
- Polyethylene succinate
- Polyethylene adipate
- Polybutylene adipate
- Polybutylene terephthalate
- Polypropylene (8x)
- Polyethylene (2x)
- Polystyrene
- PMMA
- ASA
- ABS
- PC

10 of these polymers have been assessed to develop a protocol for **melt strength**

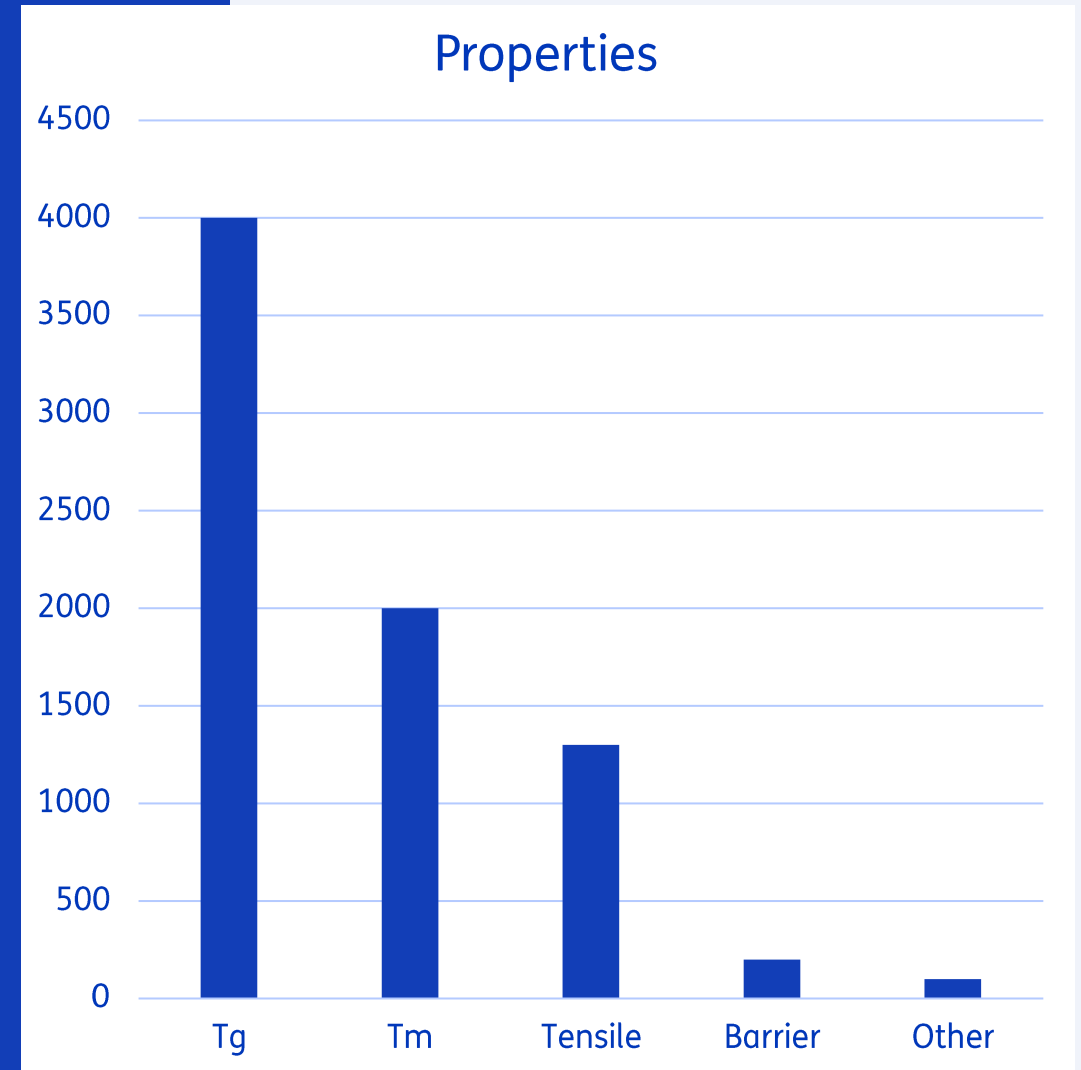
10 of these polymers have been assessed to develop a protocol for measuring **crystallization rate**

45 of these polymers have been tested for **hardness**

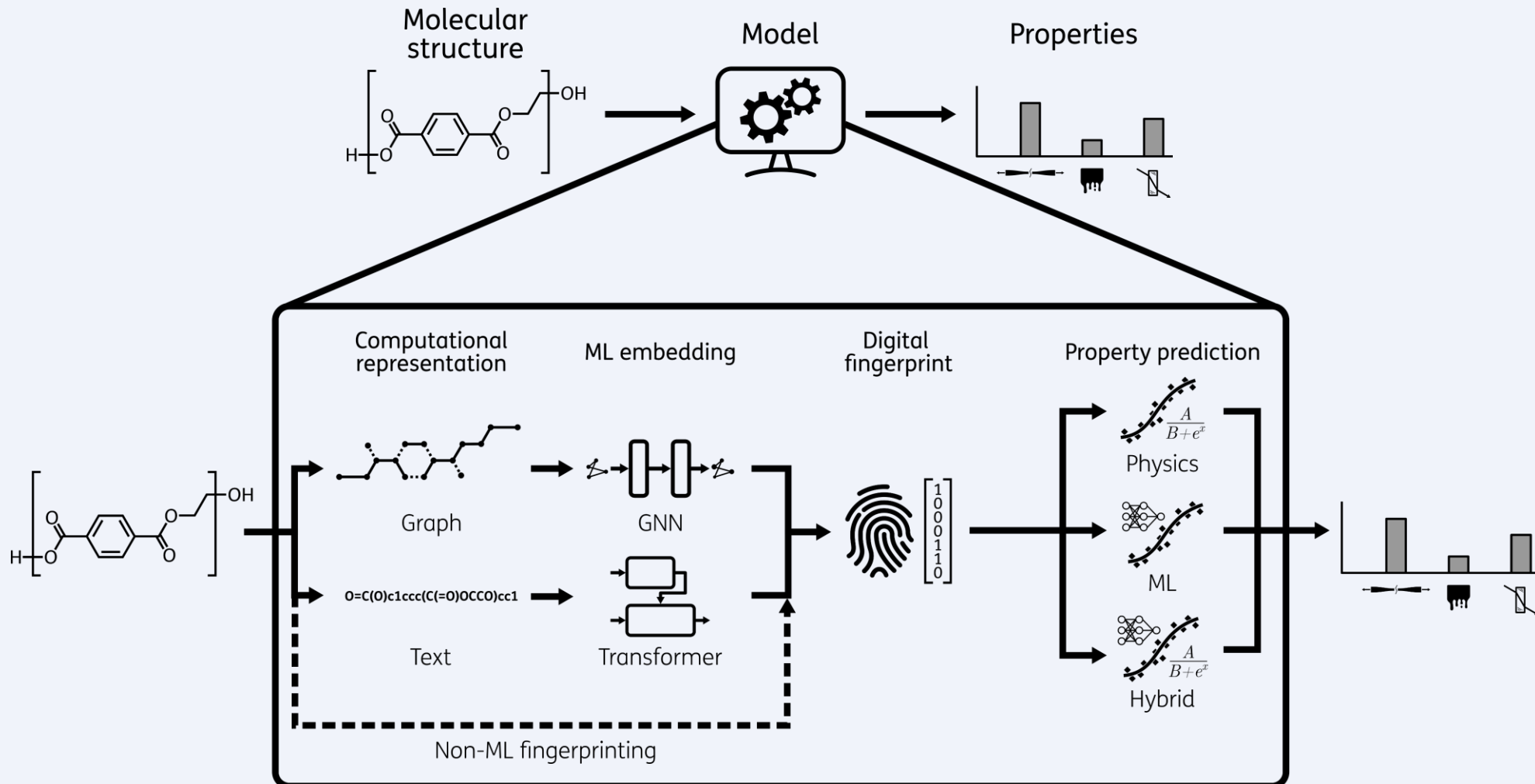
PolyScout database



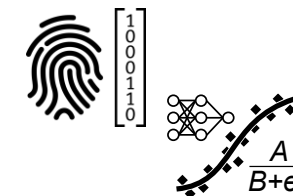
PolyScout database



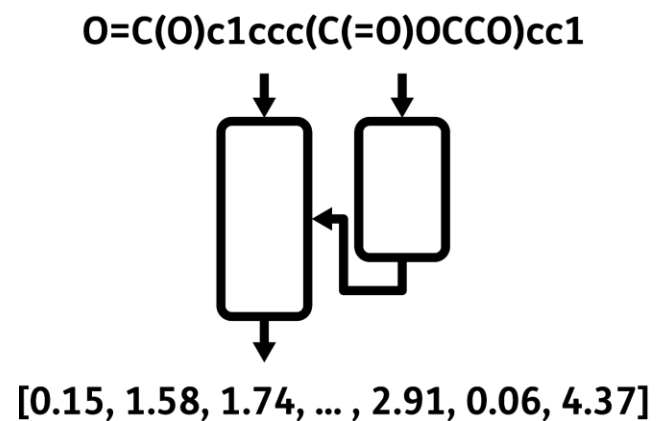
Fingerprint + Artificial Neural Network



Property prediction

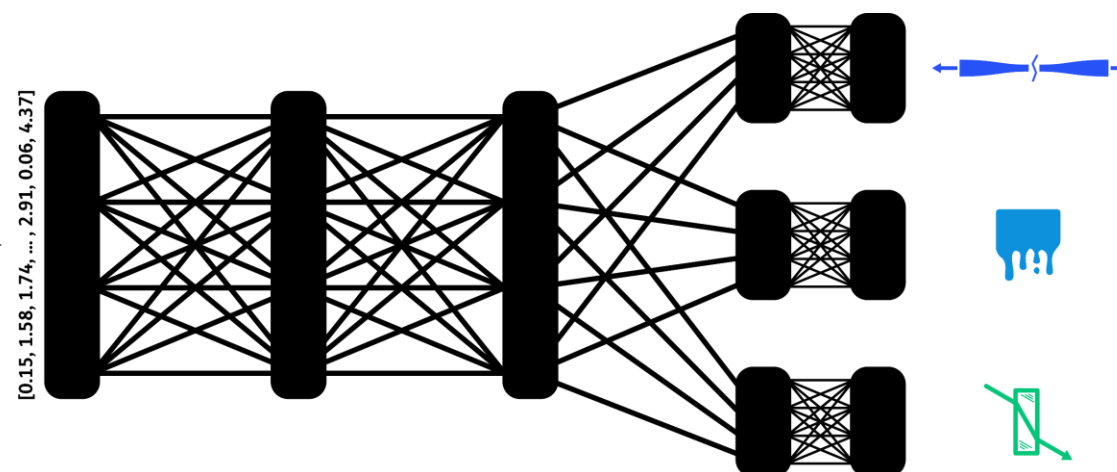


Transformer-based fingerprinting



Extended copolymer
representation in progress

Multi-task property prediction

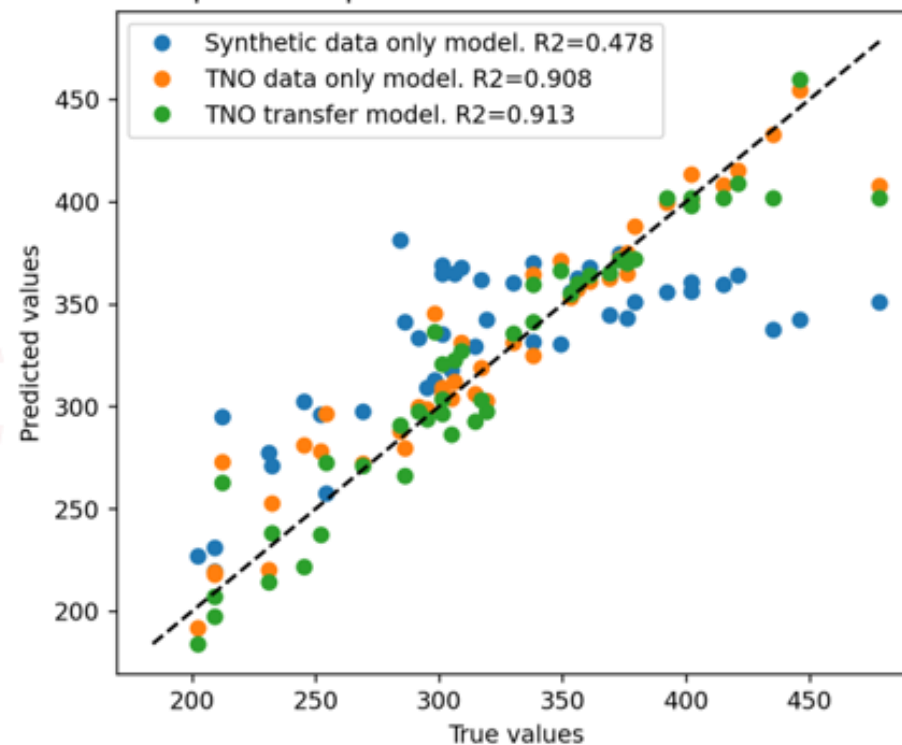


PolyScout database

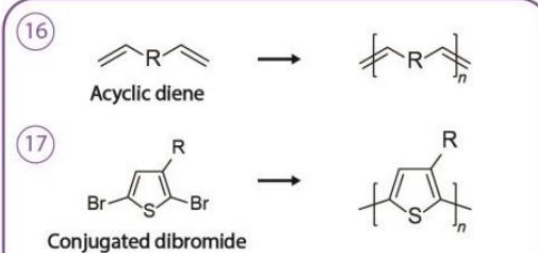
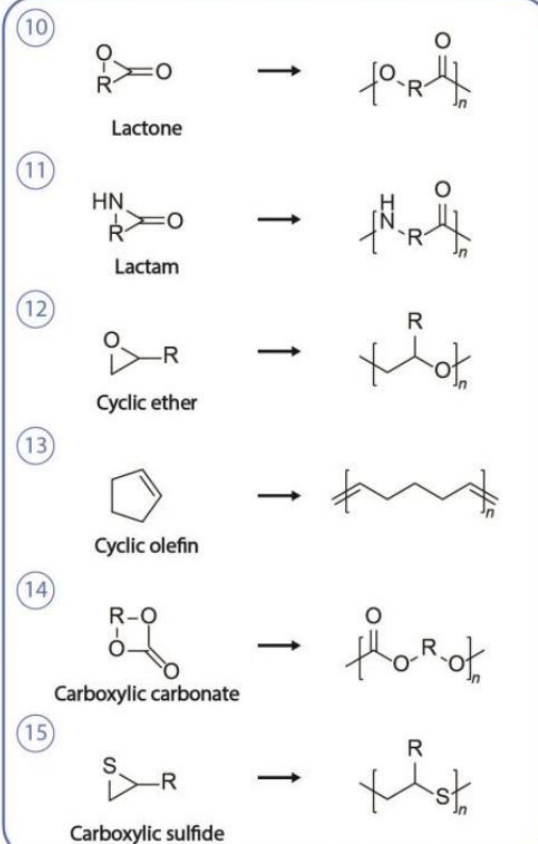
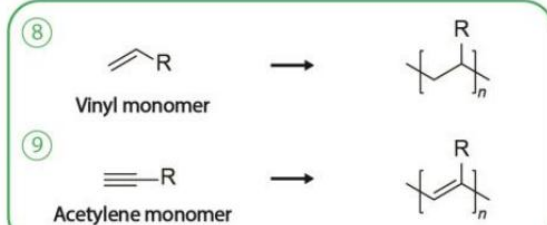
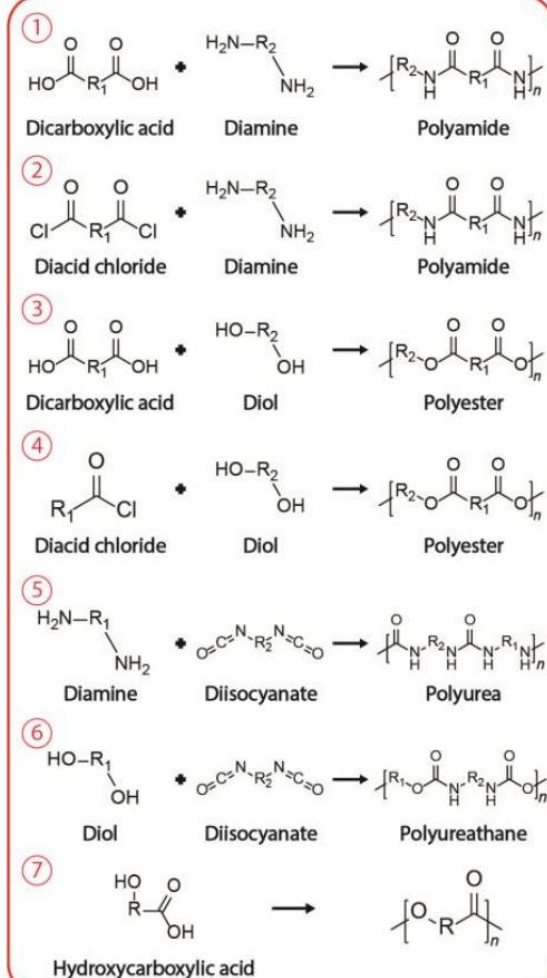


Regression plot

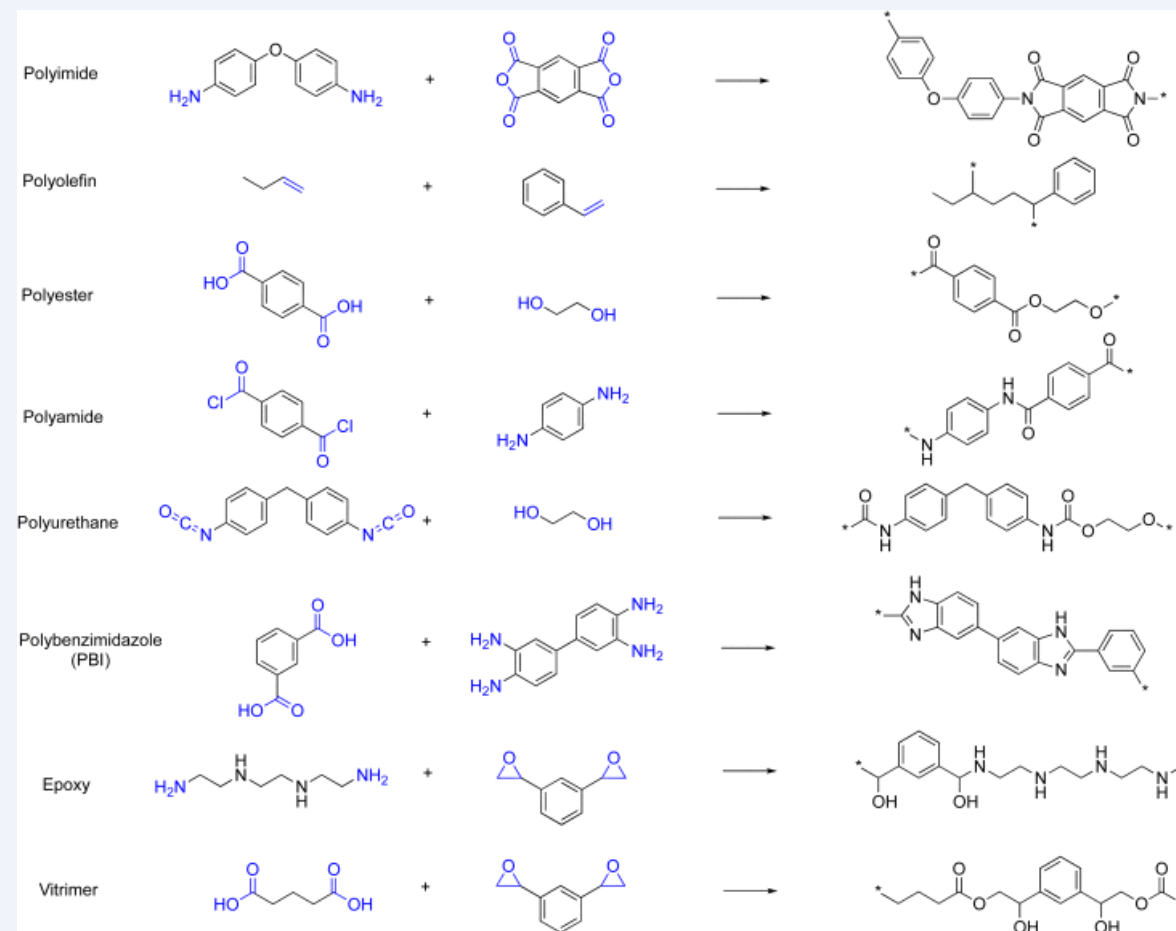
Comparison of predictions on test set vs actual values



Rule-based polymerization mechanism

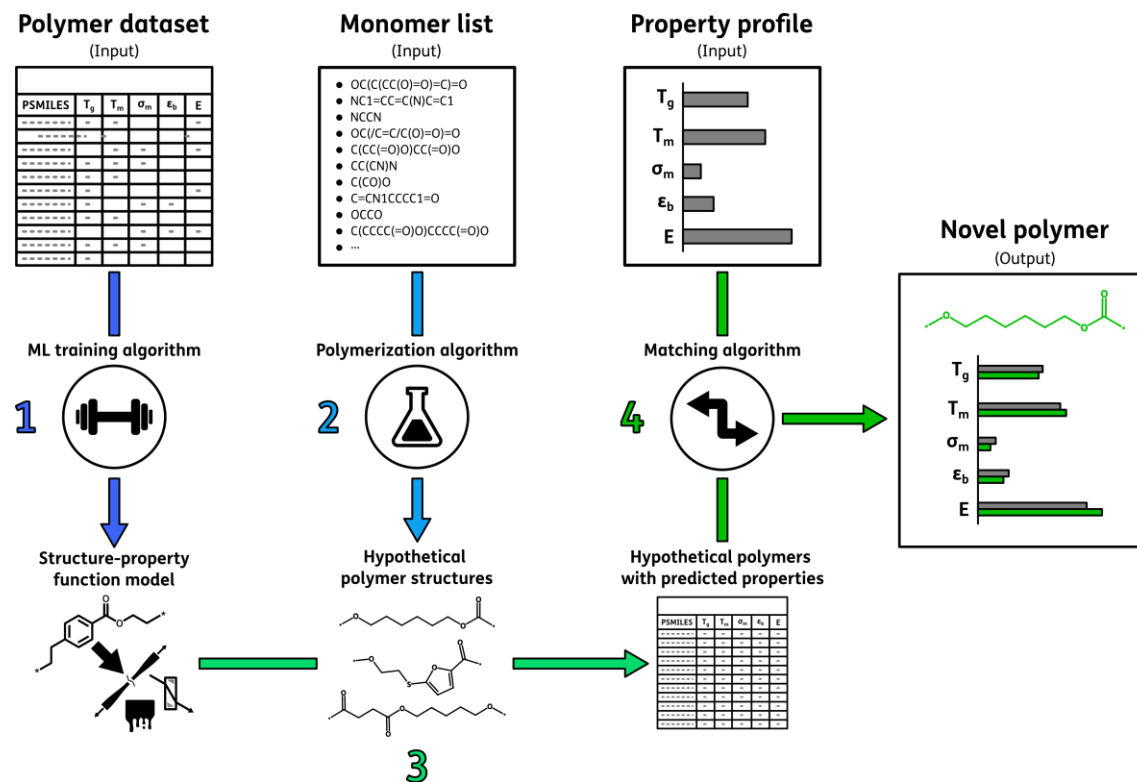


Red - step growth Blue - ring opening
Green - chain growth addition Purple - metathesis

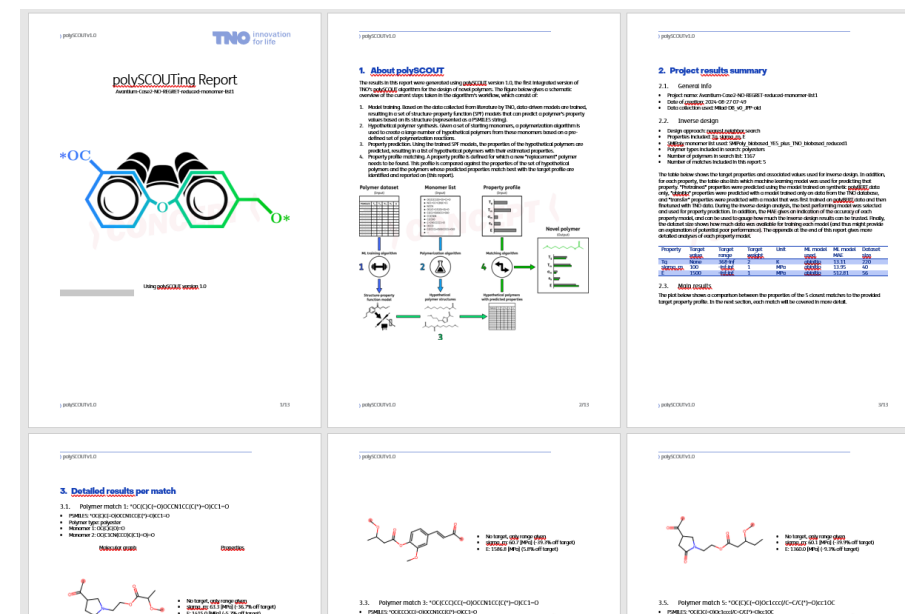


Inverse polymer design

High-throughput virtual screening approach



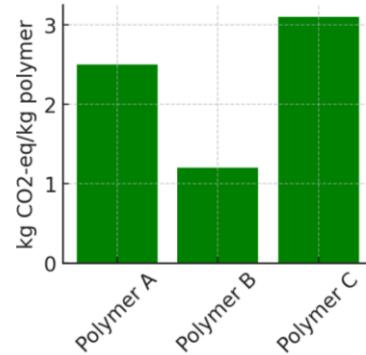
Automated reporting of results



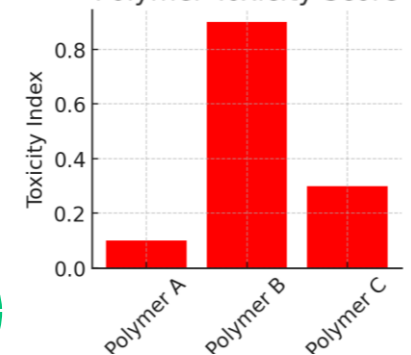
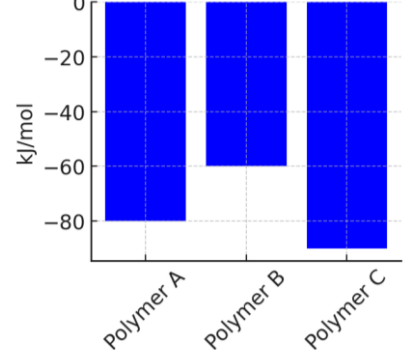
SSbD

1. **CO₂ (Life Cycle Assessment):** Global Warming Potential (GWP) estimated via machine learning models.
2. **Toxicity:** toxicity scores of different polymers, highlighting Polymer B as the most toxic.
3. **Cost (Aspen modeling):** Energy consumption for polymerization, modeled conceptually using Aspen—Polymer B is most energy-intensive.
4. **Recyclability (Enthalpy of Polymerization):** Negative enthalpy values predicted by ML show thermodynamic favorability for recyclability.
5. **Resource Availability:** Biomass feedstock availability (e.g., corn, sugarcane, cellulose, algae) shown as potential monomer sources.

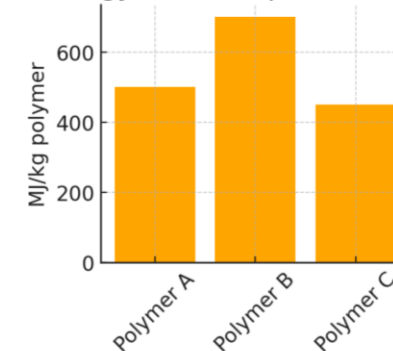
GWP from ML-based LCA



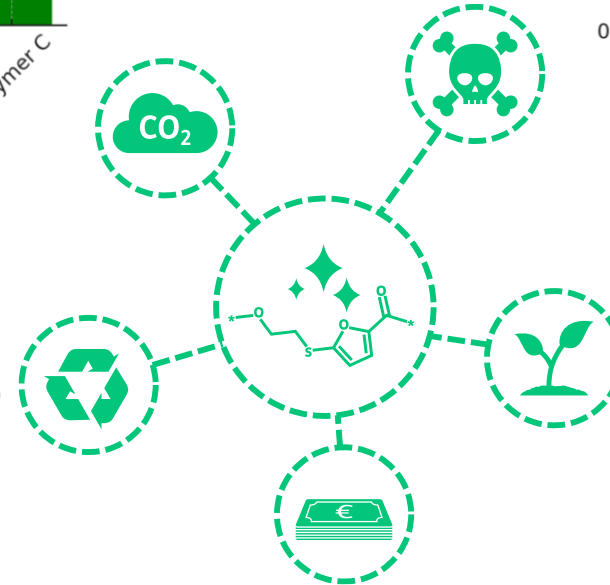
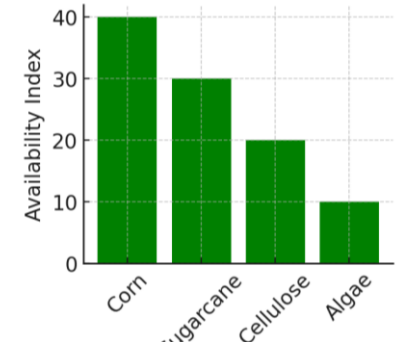
Polymer Toxicity Score

 ΔH of Polymerization (ML)

Energy Use in Aspen Simulation

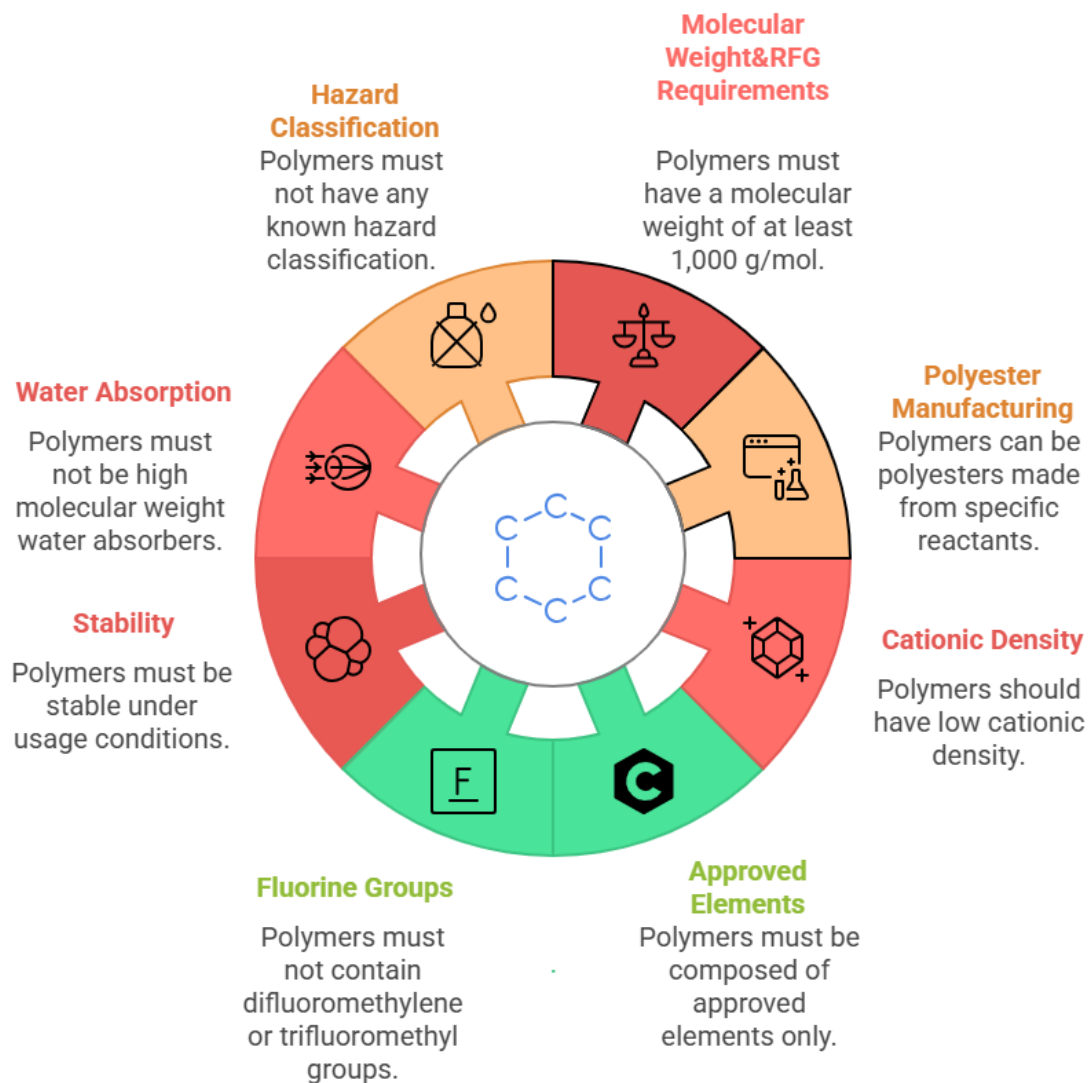


Biomass Sources for Monomers



SSbD

PLC Eligibility Criteria



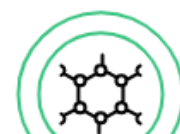
Surface Activity Determination

Assessing the impact of surface properties on toxicity



Monomer Toxicity Evaluation

Assessing the toxicity of individual chemical components



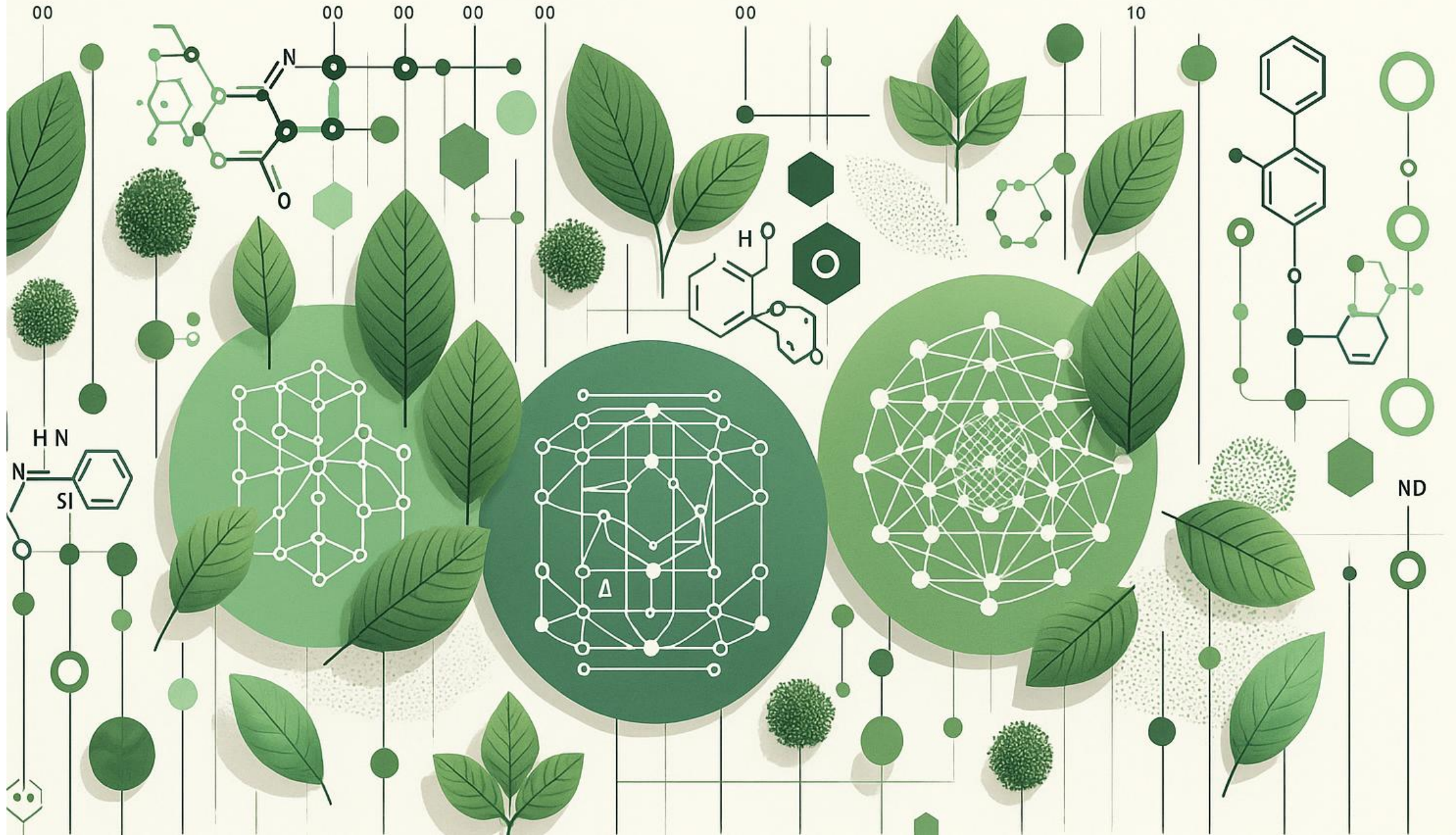
Processing Aids/Additives Toxicity

Evaluating the safety of substances used in processing



'Australian approach'

Let's collaborate to drive innovation in biobased polymers!

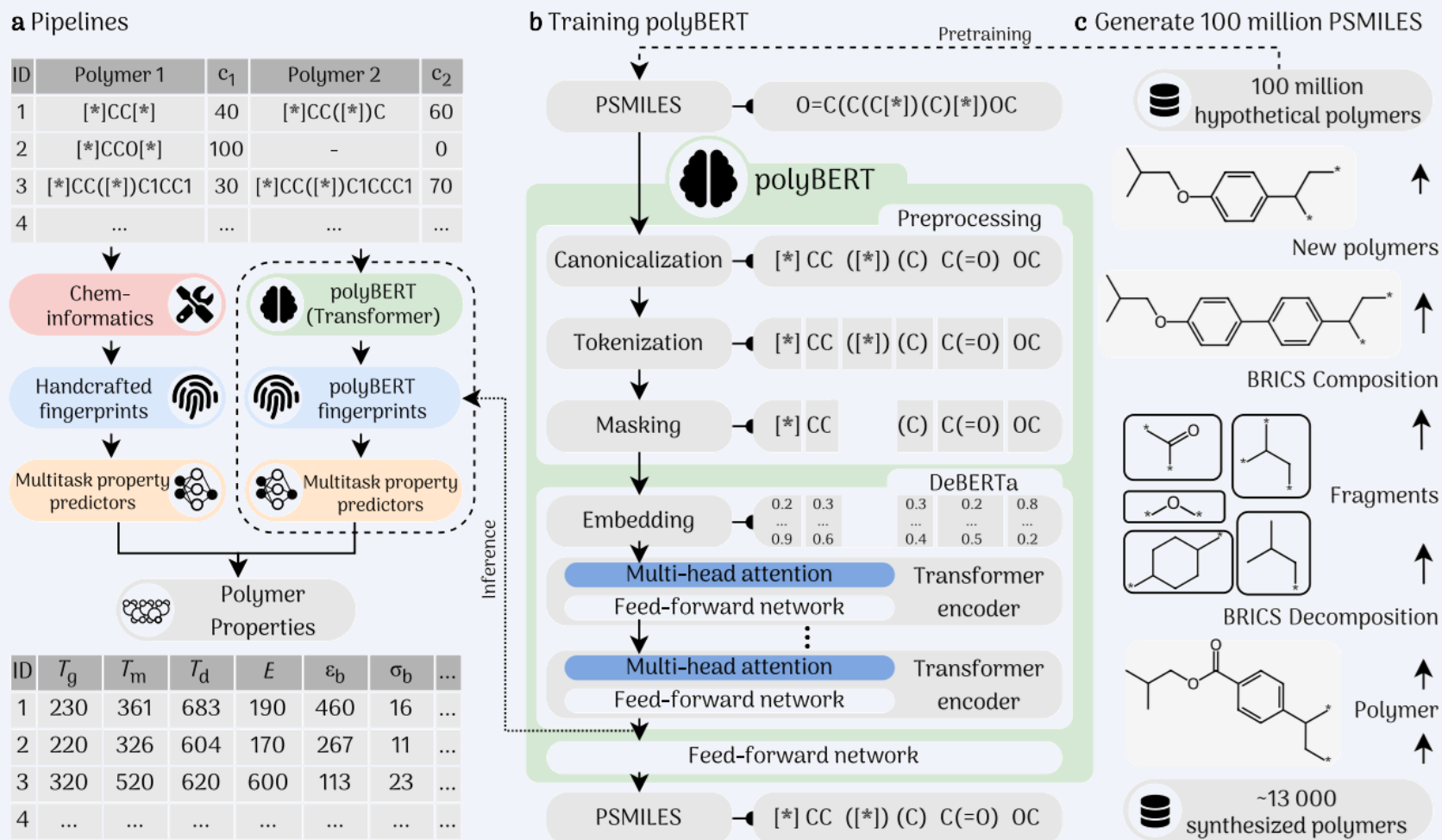


Thank you!

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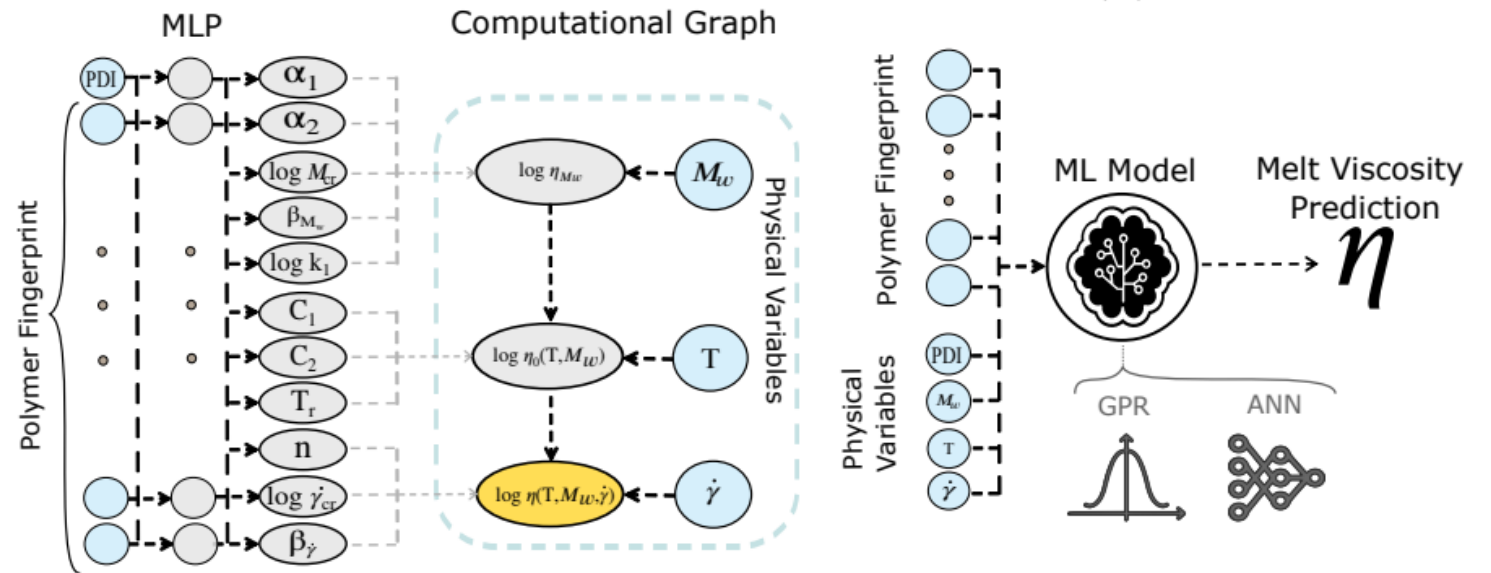
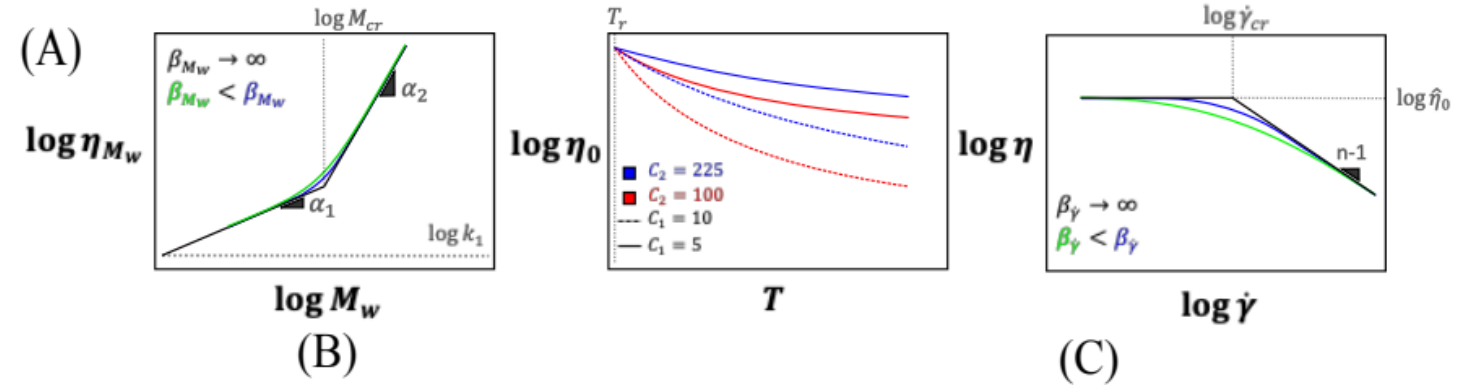
Machine learning approach



Hybrid models

- Physics-informed, physics-guided

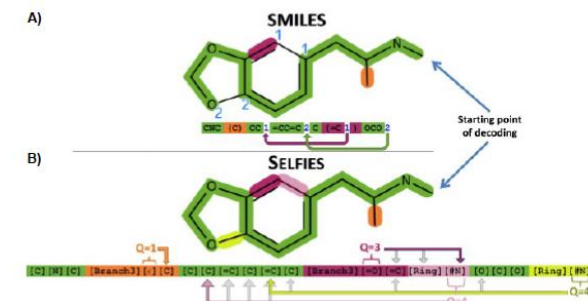
Enforced Polymer Physics Trends



Reference: 2409.05240

Polymer complexity

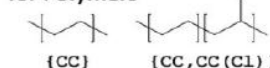
- Various representations, polyBERT uses PSMILES
- BigSMILES¹ provide better options to represent complexity of polymers (topology)



SMILES Representation
for Organic Molecules



BigSMILES Representation
for Polymers



BigSMILES Supports
a Wide Range of Structures

