# State-of-the-Art Computational Approaches for Characterizing Reaction Properties and Predicting Unknown Reactivity 

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## A $\rightarrow$ B

- To safely plan a known reaction, we need access to solid thermodynamic data (e.g., $\Delta H_{f}, S^{\circ}, C_{v}$ ) to understand and classify risks.
- This is a "known unknown" in that we know the reaction, $A \rightarrow B$, but we need values for a few unknown variables.


## $A \rightarrow$ ? $\rightarrow B ; A \rightarrow B+$ ? ; A $\rightarrow$ ?

- $\mathbf{A} \rightarrow$ ? $\rightarrow \mathrm{B}$, means that we know the net reaction, but there may be a consequential (e.g., potentially reactive) intermediate. Even if we have accurate thermodynamic data on $A / B$, neglecting the intermediate could be disastrous.
- The $\mathbf{A} \rightarrow \mathbf{B +}$ ? (unknown side-reaction) and $\mathbf{A} \rightarrow$ ? (unknown main product), problems have similar "unknown unknown" characteristics.


## $A \rightarrow B$

## TAFFI Component Increment Theory

 - This is a "known unknown" in th CTT) e reaction, $A \rightarrow B$, but we need values for a few unknown variables.
# $\mathrm{A} \rightarrow$ ? $\rightarrow \mathrm{B} ; \mathrm{A} \rightarrow \mathrm{B}+$ ? ; $\mathrm{A} \rightarrow$ ? <br>  thermodynamic data on $A / B$, neg $\mathcal{D}$ Drmediate could be disastrous. (YARP) <br> - The $\mathrm{A} \rightarrow \mathrm{B}+$ ? (unknown side-reaction) and $\mathrm{A} \rightarrow$ ? (unknown main product), problems have similar "unknown unknown" characteristics. 

## Challenges of Contemporary Group Theories

## Benson Group Theory:

- The idea is to decompose molecular properties ( $\Delta \mathrm{H}_{\mathrm{f}}, \mathrm{S}^{\circ}, \mathrm{C}_{\mathrm{v}}$ ) as the sum of "group" contributions.

From Anslyn and Dougherty's Textbook


- Group contributions are calculated based on trusted experimental or computational data, and transferability is assumed.


## Problems we want to address:

- Specificity: the definition of a "group" has never been formalized and inconsistent granularity is applied.

1) $\mathrm{C}-(\mathrm{C})(\mathrm{H})_{3} \ldots \ldots \ldots . . .2(-10.20)$
2) C -(C) $)_{3}(\mathrm{H}) \ldots \ldots \ldots \ldots . .$.
3) $\mathrm{C}-\left(\mathrm{C}_{\mathrm{B}}\right)(\mathrm{C})(\mathrm{H})_{2} \ldots \ldots \ldots \ldots-{ }^{-4.86}$
4) $\mathrm{C}_{\mathrm{B}}$-(C) $\ldots \ldots \ldots \ldots \ldots . . .$.


- Provenance: inconsistent thermodynamic data is available/used to determine group contributions.
$-5.15 \mathrm{kcal} /$ mole
$(-21.6 \mathrm{~kJ} / \mathrm{mole})$

Experimental $\Delta \mathbf{H}_{\mathrm{f}}:-5.15+/-0.34 \mathrm{kcal} / \mathrm{mol}$

- Extensibility: because of the provenance and specificity problems, it isn't possible to develop new groups in a consistent way.


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## Benson Group Theory:

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molecular properties $\left(\Delta H_{f}, S^{\circ}, C_{v}\right)$ as the
sum of "group" contributions.

> Can we circumvent the provenance and extensibility challenges using the throughput and accuracy of modern quantum chemistry?

Zhao, Q.; Savoie, B. M.; Enthalpy of Formation Prediction via a fully Self-Consistent
Component Increment Theory. J. Chem. Info. Model. 2020, 60, 2199-2207

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## TAFFI Component Increment Theory (TCIT)

## The fundamental idea

- Systematize component-definitions and model compound selection with rigorous graph-based typing.


## TCIT is a component theory <br> (2-bond specific)



Topology Automated Force Field Interactions

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Adjacency matrix for PEDOT

Seo, B.; Lin, Z.-Y.; Zhao, Q.; Webb, M. A.; Savoie, B. M. Topology Automated Force-Field Interactions (TAFFI): A Framework for Developing Transferable Force Fields. J. Chem. Inf. Model. 2021, 61 (10), 5013-5027. https://doi.org/10.1021/acs.jcim.1c00491.

P2SAC Publications

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P2SAC Publications

Adjacency matrix for PEDOT
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Topology Automated Force Field Interactions


P2SAC
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Adjacency PEDOT

## TAFFI Component Increment Theory (TCIT)

## The fundamental idea

TCIT is a component theory
(2-bond specific)

- Systematize component-definitions and model compound selection with rigorous graph-based typing.
- Two-bond specificity should improve both the accuracy and transferability of the resulting components.
- Parameterizing a component model would not be feasible with only experimental data.

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P2SAC
Publications


Adjacency matrix for PEDOT monomer

## Graphical Decomposition of Model Compounds

<br>1-hydroxy-pent-2-ene-2-one

## Graphical Decomposition of Model Compounds

identify components*


1-hydroxy-pent-2-ene-2-one

How will we select molecules for parameterizing TCIT components?

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## How will we select molecules for parameterizing TCIT components?

Recursively generate smallest acyclic model compounds

Resolve dependencies




## Graphical Decomposition of Model Compounds



## Graphical Decomposition of Model Compounds



## Graphical Decomposition of Model Compounds

Have we solved the specificity problem?
All components are unique out to a graph depth of two, no exceptions.

Have we solved the provenance problem?
All $\Delta \mathrm{H}_{\mathrm{f}}$ data is calculated at the G4 composite level, no exceptions.

Have we solved the extensibility problem?
Model compounds exist for all conceivable components, no exceptions.

## Benchmarking $\Delta \mathrm{H}_{\mathrm{f}, \mathrm{gas}}$ Predictions Against the PNK Dataset

- Initial benchmarking set consists of $\sim 1100$ linear $\mathrm{C}, \mathrm{H}$, and O containing compounds from $\mathrm{PNK}^{1}$
(1) J. B. Pedley, R. D. Naylor, S. P. Kirby "Thermochemical Data of Organic Compounds" $2^{\text {nd }}$ ed. 1986
- PNK is a core dataset for fitting Benson groups
- ~600 PNK compounds are small enough for G4 calculations and comparison with experiment.


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- ~150 PNK compounds are large enough for direct G4 calculation and comparison with TCIT.

150 medium compounds from PNK


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- ~500 PNK compounds are large enough to evaluate the predictive accuracy of the increment theories.


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> TCIT shows comparable performance to BGIT/CHETAH but is derived exclusively from extensible G4 data.

## Extension to Ring-Containing Molecules

- Ring-containing molecules have additional strain and/or conjugation corrections that exacerbate the extensibility issues of Benson Theory.
- In TCIT we are addressing this through chemically specific ring corrections that account for differences in substitution pattern and topology:

1. Decompose ring into acyclic model compounds:


Technical Developments


Method 1: Use $\mathrm{RC}_{1}$ based model parameterized to G4 data.
Method 2: Use graph-NN to predict $\mathrm{RC}_{0}-\mathrm{RC}_{2}$
2. Add ring correction (RC) to final prediction:

| Technical Developments |
| :--- |
| Method 1: Use $\mathrm{RC}_{1}$ based model parameterized <br> to G4 data. <br> Method 2: Use graph-NN to predict $\mathrm{RC}_{0}-\mathrm{RC}_{2}$ |
| : DC Model Compounds |

$$
\mathrm{RC}=H_{f}(\text { ring })-H_{f}()-H_{f}()-H_{f}()-H_{f}()-H_{f}()-H_{f}()
$$

## Benchmarking Ring-Correction Performance

(a) G4 errors are marginally larger for ring-containing compounds but still very accurate
(b) The neural-network based ringcorrection exhibits excellent reproduction of the G4 predictions (MSE: ~3kJ/mol; MAE: ~8 kJ/mol).
(c) TCIT is completely transferable to new testing compounds that are experimentally characterized. Errors are consistent with G4 comparison
(d) The TCIT-R2 model outperforms BGIT on the large molecule benchmark while being extensible. Significantly, these compounds are within BGIT's training data.
~120 ring-containing compounds from PNK (excluding training)


BGIT cannot make predictions for $\mathbf{\sim 2 \%}$ of PNK compounds

## Benchmarking Ring-Correction Performance

## Breaking down distinct subsets:

- Conjugated systems are challenging to accurately predict with a local ring correction.
- BGIT has excellent performance on benzene rings due to the prevalence of experimental data, but poor performance on novel rings.
- The ML ring-correction shows the strongest overall performance. This strategy could also be used to generically correct for long-range conjugation effects.
~120 ring-containing compounds from PNK (excluding training)


TCIT-ML
BGIT cannot make predictions for $\mathbf{\sim} \mathbf{2 \%}$ of PNK compounds

## TCIT Extension to Other Properties and Phases

Condensed Phases: The condensed-phase and gas-phase standard enthalpies of formation differ by the heats of sublimation and vaporization ${ }^{[1]}$ :

$$
\begin{aligned}
& \Delta_{\mathrm{f}} H_{(\mathrm{s})}^{\circ}=\Delta_{\mathrm{f}} H_{(\mathrm{g})}^{\circ}-\Delta_{\mathrm{sub}} H^{\circ} \\
& \Delta_{\mathrm{f}} H_{(\ell)}^{\circ}=\Delta_{\mathrm{f}} H_{(\mathrm{g})}^{\circ}-\Delta_{\mathrm{vap}} H^{\circ}
\end{aligned}
$$

We have implemented group contribution models for heat of vaporization ${ }^{[2]}$ and sublimation ${ }^{[3]}$, respectively. The group assignments and group values associated with these models have been automated within the context of TCIT.

Standard Molar Entropy ( $\mathbf{S}^{\boldsymbol{o}}$ ) and heat capacity ( $\mathbf{C}_{\mathbf{v}}$ ): The molar entropies and constant volume heat capacities are accessible from quantum chemistry using the harmonic oscillator approximation for the molecular partition function and corrections based on the number of rotatable bonds ( $\mathrm{N}_{\text {rot }}$ ) and molecular symmetry:

$$
S^{\circ}=\left\langle S_{\mathrm{harm}}^{\circ}\right\rangle+R N_{\mathrm{rot}}+R \log \sigma \quad C_{\mathrm{v}}=\left\langle C_{\mathrm{v}, \mathrm{harm}}\right\rangle+\alpha N_{\mathrm{rot}}+\beta
$$

[1] Murray, J.S., Brinck, T. and Politzer, P., 1996. Chemical physics, 204, 289-299.
[2] Pankow, J.F. and Asher, W.E., 2008. Atmospheric Chemistry and Physics.
[3] Bagheri, M.; Gandomi, A. H.; Golbraikh, A. 2012, Thermochim. Acta, 543, 96-106
<.> Indicates conformational averaging
$R$ : ideal gas constant
$\sigma$ : symmetry number
$\alpha, \beta$ : regressed constants

## Benchmarking Condensed Phase $\Delta \mathrm{H}_{\mathrm{f}}$ Predictions



- Testing set includes both linear and cyclic compounds with number of heavy atoms varying from 1 to 30 .
- Low MSE indicates no systematic bias, larger absolute errors result from the quality of the $\Delta \mathrm{H}_{\text {vap }}$ and $\Delta \mathrm{H}_{\text {sub }}$ models.
[1] Linstrom, P.J. and Mallard, W.G., 2001. Journal of Chemical \& Engineering Data, 46(5), pp.1059-1063.


## Benchmarking TCIT $\mathrm{S}^{\circ}$ and $\mathrm{C}_{\mathrm{v}}$ Predictions

(a) G4/TCIT $\mathrm{S}^{\circ}$ comparison for 314 medium sized molecules.
(b) G4/TCIT $\mathrm{S}^{\circ}$ comparison for 314 medium sized molecules.
(c) TCIT $\mathrm{S}^{\circ}$ comparison for 439 large molecules from NIST[1]
(d) TCIT heat capacity comparison for 904 large molecules from NIST ${ }^{[1]}$

- The TCIT errors are consistent with error propagation of G4:exp and TCIT:G4 errors.

TCIT now supports $\mathrm{S}^{\circ}$ and $\mathrm{C}_{\mathrm{v}}$ predictions with accuracies comparable to G4 model chemistry.

(C) $850-2$ TCIT


[1] Linstrom, P.J. and Mallard, W.G., 2001. Journal of Chemical \& Engineering Data, 46(5), pp.1059-1063.

## Benchmarking TCIT $\mathrm{S}^{\circ}$ and $\mathrm{C}_{\mathrm{v}}$ Predictions



[^0]
## How Many Components are Possible?

Prediction target:


1-hydroxy-pent-2-ene-2-one
Gen 4:


Gen 3:


Gen 2:



Gen 1



- OH

Gen 0: $=$
©

We database all model compounds and components for reuse.

Over the past two years, we have parameterized new components in response to distinct project needs (many from P2SAC Pharma Members)

## Current Database:

- ~35k distinct components for $\Delta H_{f}$ relevant to organic chemistry - ~35k distinct G4 calculations on organic molecules.
- ~450 distinct ring corrections


## How Many Components are Possible?

How many components are required to predict the $\Delta H_{f}$ of all (physically relevant) organic molecules?


How many P2SAC funding periods would it take to make a "complete"/gapless component theory?

Gen 0:

## Treating PubChem as a Model of Organic Chemical Space

PubChem is a repository of chemical properties that contains many millions of organic species ranging from small molecules to oligonucleotides.

We recently started mining PubChem's H,C,N, and O containing molecules for distinct components and the model compounds necessary to predict $\Delta H_{f}$


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The derivative plot shows that TCIT initially generates $\sim 2$ new components per molecule, but by the end of the sampling ~100 molecules need to be sampled to find a new component.


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We recently started mining PubChem's $\mathrm{H}, \mathrm{C}, \mathrm{N}$, and O containing molecules for distinct components and the model compounds necessary to predict $\Delta H_{f}$

The derivative plot shows that TCIT initially generates ~2 new components per molecule, but by the end of the sampling ~100 molecules need to be sampled to find a new component.

New model compounds

## Treating PubChem as a Model of Organic Chemical Space

In the past six months we have generated all training data necessary to make predictions on all $\mathrm{N}, \mathrm{H}, \mathrm{O}$, and C-containing molecules in pubchem. This is the largest repository of G4 calculations on large molecules in the world.

It is foreseeable that we could complete all B, F, CI, S, and P containing structures over the next few years.

## A $\rightarrow$ B

- To safely plan a known reaction, we need access to solid thermodynamic data (e.g., $\Delta H_{f}, S^{\circ}, C_{v}$ ) to understand and classify risks.
- This is a "known unknown" in that we know the reaction, $A \rightarrow B$, but we need values for a few unknown variables.


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- $\mathbf{A} \rightarrow$ ? $\rightarrow \mathrm{B}$, means that we know the net reaction, but there may be a consequential (e.g., potentially reactive) intermediate. Even if we have accurate thermodynamic data on $A / B$, neglecting the intermediate could be disastrous.
- The $\mathbf{A} \rightarrow \mathbf{B +}$ ? (unknown side-reaction) and $\mathbf{A} \rightarrow$ ? (unknown main product), problems have similar "unknown unknown" characteristics.
$\mathrm{A} \rightarrow \mathrm{B}:$ When we know the reactants and products, mature quantum chemistry tools exist to characterize transition states and establish pathways

A $\boldsymbol{\rightarrow}$ ? : For degradation reactions, plausible reactions are often unknown.

## The Reaction Prediction Problem

A $\rightarrow$ B: When we know the reactants and products, mature quantum chemistry tools exist to characterize transition states and establish pathways

A $\boldsymbol{\rightarrow}$ ? : For degradation reactions, plausible reactions are often unknown.

Thermal, $\mathrm{pH}, \mathrm{h} v, \mathrm{O}_{2}$,


3-hydroperoxypropanal

## Yet Another Reaction Program (YARP)

Idea: Turn the $\mathbf{A} \boldsymbol{\rightarrow}$ ? problem into tractable (and parallelizable) $\mathbf{A} \rightarrow \mathbf{B}$ problems.

## Observations:

- Product enumeration is easier than transition state enumeration.
- Transition state algorithms for $A \rightarrow B$ problems are mature. Let the TS algorithm identify physical reactions.
- Recent developments in semi-empirical models and ML create opportunities.
- If you are fast enough, you can brute force the $\boldsymbol{A} \boldsymbol{\rightarrow}$ ? problem.



## Converting Reactions into a Machine-Readable Grammar

Bond-Electron Matrix Formalism: matrix representation of molecules with bond order indicated in off-diagonal elements and lone electrons along the diagonal.

Ugi, I. et al. "New Applications of Computers in Chemistry." Angew. Chem. 1979, 18 (2), 111-123.

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This is essentially a way of converting arrow pushing into something we can automate and interpret by a program

## YARP: Elementary Reaction Step(s)

For full-octet neutral organic molecules, "break 2 bonds form 2 bonds" (b2f2) is the simplest ERS that yields non-trivial closed-shell neutral products.


Example with concomitant e-transfer


Note: for each pair of broken bonds two distinct reactions are possible

YARP supports a more general suite of ERS(s), but b2f2 strikes a useful balance that scales as $\mathrm{N}^{2}$ with reactant size and generates b1 and b2f1 type products for free.

## YARP: Elementary Reaction Step(s)

For full-octet neutral organic molecules, "break 2 bonds form 2 bonds" (b2f2)
is the simplest ERS that yields non-trivial closed-shell neutral products.

## All b3f3 and b4f4 products are b2f2 decomposable

This means that using only b2f2 won't miss any products, but it will potentially miss important transition states (i.e., by predicting a sequential mechanism when a concerted mechanism is favored)

OH
distinct reactions are possible

YARP supports a more general suite of ERS(s), but b2f2 strikes a useful balance that scales as $\mathrm{N}^{2}$ with reactant size and generates b1 and b2f1 type products for free.

## YARP: Pseudo-1D Transition State Searches

## Searching for saddle points in $3 \mathrm{~N}-6$ space is expensive.



Double-Ended Searches: Using the product and reactant geometries as a constraint, the search can be recast as an effective 1-D search about the connecting coordinate.

## Many Flavors:

- (Growing, Freezing) String methods
- Nudged-Elastic band methods

Peters, B.; Heyden, A.; Bell, A. T.; Chakraborty, A. J. Chem. Phys. 2004, 120 (17), 7877-7886.

We like strings because they avoid the bad initial pathway problems of bands.

## Three Sources of Error in TS Searches



## Three Sources of Error in TS Searches

## Not every reactant-product pair has a unique TS

This means that a common failure mode for unbiased searches is that your transition state finding algorithm simply fails. This makes arbitrarily attempting ill-conditioned reactions a costly mistake.

Reaction Coordinate
Reaction Coorainate
Reaction coorainate

Failure to localize a
transition state for a given
$A \rightarrow B$ reaction

Failure to localize the
kinetically relevant TS for
a given reaction

Failure to localize an
intended TS for a given reaction

## Testing YARP on a Unimolecular Decomposition Problem

## 

The 3-hydroperoxypropanal reaction network out to b4f4 was recently published as a benchmark for 5 reaction discovery methods.

Grambow, C. A, Suleimanov, Y. V. et al. J. Am. Chem. Soc. 2018, 140 (3), 1035-1048.
$\underline{\mathrm{H}}_{2} \mathrm{O} \pm$ malondialdehyde channels:

Biradical products including carbenes and the Criegee intermediates:


Zwitterionic structures:
R18,19,26

R20

## R21

R22

## R22


$\ominus \mathrm{O}-\stackrel{\oplus}{\mathrm{O}} \mathrm{H}_{2}=\mathrm{CO}$
P24,25


R27

$\mathrm{H}_{2}$

Channels with three products except zwitterionic structures
$\mathrm{R} 28 \quad \mathrm{H}_{2} \mathrm{C}=\mathrm{C}=\mathrm{O} \quad \mathrm{H}_{2} \mathrm{C}=\mathrm{O} \quad \mathrm{H}_{2} \mathrm{O}$


CO $\mathrm{H}_{2}$
$\mathrm{R} 30 \mathrm{HO}-\mathrm{OH}=\mathrm{CO}$
R31


Channels with cyclic products:


R33,34


Stable (not radical or zwitterionic) unimolecular noncyclic channels:


R55
R56

$\underline{H}_{2}$ elimination channels:



R62

$\mathrm{H}_{2}$

Non-malondialdehyde $\underline{H}_{2} \mathbf{O}$ elimination channel:

R63
 $\mathrm{H}_{2} \mathrm{O}$
$\mathrm{CH}_{2}-\mathrm{CH}_{2}$ or $\mathrm{CH}_{2}-\mathrm{CHO}$ bond breaking and forming two non-cyclic products: R64 $\mathrm{H}_{2} \mathrm{C}=\mathrm{O}$ $\mathrm{R} 65,66 \quad \mathrm{H}_{2} \mathrm{C}=\mathrm{C}=\mathrm{O}$ $\mathrm{R}_{67,68,73} \mathrm{HO}^{\mathrm{O}} \mathrm{H}_{2} \mathrm{C}=\mathrm{O}$ R69*


HOOH elimination channels
R74

## Testing YARP on a Unimolecular Decomposition Problem

## $\mathrm{H}_{2} \mathrm{O} \pm$ malondialdehyde channels <br> Biradical products including <br> carbenes and the Criegee intermediates:

## 55 distinct products (neglecting charged species)



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## 562 reactions attempted

## anneis with three products

75 intended reactions discovered

None of the methods found all of
Zwitterionic structures: the products of the others

flimination channels:

## 3-Hydroperoxypropanal - Reaction Network

We used YARP to recursively elucidate the 3hydroperoxypropanal unimolecular thermal degradation network for comparison with Grambow et al.

YARP finds all known products of this thermal decomposition network, as well as new products (77), and new reactions (157).


## 3-Hydroperoxypropanal - Reaction Network

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 of this thermal decomposition network, as well as new products (77), and new reactions (157).

## Predicting More (Reactions) with Less (Cost)




Constructing the whole network required 8364 DFT gradient calls for YARP compared with $\mathbf{7 5 6}, 227$ for the earlier benchmark (100-fold reduction)
Average success and intended rates for YARP are 81.4\% and 41.1\%, respectively, compared with $38 \%$ and $4 \%$, in the earlier benchmark.

## What Happens First?

Jensen, R. K.; Korcek, S.; Mahoney, L. R.; Zinbo, M. JACS 1979, 101, 7574

## The Korcek Mechanism

According to YARP, this is the lowest barrier degradation product.



Validated 30 years later by Green and Truhlar:
Jalan, A.; Alecu, I. M.; Meana-Pañeda, R.; Aguilera-Iparraguirre, J.; Yang, K. R.; Merchant, S. S.; Truhlar, D. G.; Green, W. H. JACS 2013, 135 (30), 11100-11114.

## Recent A $\rightarrow$ ? and A $\rightarrow$ ? $\rightarrow$ B Case Studies



Figure 1. Proposed pathways in literature from glucose to HMF, namely the fructose path (green), 3-DG paths (black and black dotted), and direct path (red). The molecules are indicated by numbers and some key molecules are named as follows: 1. D-glucose; 2. D-fructose; 3. D-fructofuranose; 6. 5-hydroxymethylfurfural (5-HMF); 7. 3-deoxyglucos-2-ene (3-DGE); 8. 3-deoxyglucosone (3-DG); and 10. hex-1-ene-1,2,3,4,5,6-hexaol (enol form of glucose).

## Glucose Pyrolysis Network from YARP Exploration

To perform a deep network exploration, we've implemented a modified version of Dijkstra's algorithm


At each iteration:
(1) all b2f2 reactions are explored for active nodes.
(2) Active nodes are determined by the minimum barrier to a given product (with a window)
(3) Water catalyzed reactions are considered for all H transfers

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Depth 1: $\square$
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## Glucose Pyrolysis Network from YARP Exploration

By following the kinetically favorable pathways, YARP spontaneously discovers all major pyrolysis products.
(1) all b2f2 reactions are explored for active nodes

Pathways to minor products can also be found using backward searches.

## A Recent A $\rightarrow$ ? $\rightarrow$ B Case Study

PBDF is a new record breaking $n$-type semi-conducting plastic

 $\xrightarrow{\text { reduction }}$

in situ doping
Small molecule studies show a significant synthetic side product


## Pathways identified by YARP



## Outlook and Acknowledgements

Students: Qiyuan Zhao, Tyler Pasut

## State-of-the-art:

- The accurate calculation of thermodynamic properties has become routine in many scenarios. Major opportunities lie in automation, systemization, and low-cost models.
- Practical solutions to the $A \rightarrow$ ? $\rightarrow B$, $A \rightarrow B+$ ?, and $A \rightarrow$ ? problems are now available. We envision black-box tools for non-experts in the near future that will assist in hypothesis generation and potentially reactivity screening.

- P2SAC and ONR for funding.
- Ray Mentzer (Purdue)
- Katherine Young (Purdue UG)



[^0]:    [1] Linstrom, P.J. and Mallard, W.G., 2001. Journal of Chemical \& Engineering Data, 46(5), pp.1059-1063

