

Dana Research Group

Fundamental and Applied Chemical Kinetics

Quantitative *in-silico* kinetic predictions of API degradation

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Figure from AIDS Healthcare Foundation

Acknowledgments

Oscar Wu, MIT

May Cohen, Technion

Duminda Ranasinghe, MIT

Prof. William H. Green, MIT

Jason Mustakis, Pfizer

Frank C. Pickard, Pfizer

Gregory Sluggett, Pfizer

Geoffrey Wood, Pfizer

Todd Zelesky, Pfizer

All RMG, ARC, and T3's developers



The grand question of this talk:

Can we quantitatively predict

the degradation of API molecules?

Our Goal:

Develop an automated self-improving software to predict and explain

free-radical oxidative stability and model solution degradation kinetics of APIs

Talk Outline

Methodology

• The Radical "Soup" in API Stress Testing

• Case Studies

• Outlook and Perspective

Overarching goal:

Predict the time-evolution of a chemical system given known initial conditions.

John von Neumann famously said:

With four parameters I can fit an elephant, and with five I can make him wiggle his trunk.



We are not fitting, but rather predicting.

5 J. Mayer, K. Khairy, J. Howard, "Drawing an elephant with four complex parameters", *Am. J. Phys.* **2010**, 78, 648.

Why do we want predictive chemistry?

Predictive Science:

Accurate predictions **show we understand** the system, not just waving hands.

Understand important systems, esp. if not practical to perform experiments.

Predictive Design and System Optimization:

Develop **new products and processes** on the computer rather than trial and error.

Chemical Discovery:

Can we **predict new types of reactions**, never discovered before?

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Required Parameters

Luckily, computational chemistry

has already advanced to the stage where it has quantitative * predictive capability for reaction rates of complex species

Experimentally Available **Parameters**





Reaction Mechanism Generator



What do we need to generate a kinetic model?

all (significant?) intermediate species

all (significant?) reactions

all reaction rate coefficients

8 M. Liu, A. Grinberg Dana et al., "Reaction Mechanism Generator v3.0: Advances in Automatic Mechanism Generation", J. Chem. Inf. Model 2021, 61(6), 2686.































Where are the data (kinetics and thermochemistry) coming from?





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Existing "libraries" in RMG's database:

Compiled literature values, or own computations / measurements

Estimations:

Group- / Tree- / ML- based data estimations

Ab-initio (e.g., DFT / WF):

Complementary quantum chemical computations for selected model parameters

Need high-quality automated ab-initio thermo-kinetic computations

Introducing the **Automated Rate Calculator** (ARC) software tool:



ARC's mission is to provide the community a well-documented, extensible

codebase for automatically calculating species thermochemistry and

high-pressure limit reaction rate coefficients.

Structure discovery	QM calcs	Trouble- shoot	Statmech	Processing
3D conf. search	Geometry opt.	Convergence	Currently:	Visualizations
TS Search:	Frequencies Single-point E	Max cycles	- Arkane	RMG libraries
- GCN ^[1]	Torsional scans	Disk space	Future:	Parity plots
- KinBot ^[3]	IRC Orbitals	Memory	- Mesmer	Comparisons
- Heuristics - GSM ^[4] (xTB)	BDEs	Software-specific		

[1] L. Pattanaik et al., Phys. Chem. Chem. Phys. 2020, 22, 23618.[2] P.L. Bhoorasingh et al., J. Phys. Chem. A 2017, 121, 6896.

[3] R. Van de Vijver, J. Zador, Comp. Phys. Comm. 2020, 248, 106947.[4] P.M. Zimmerman, J. Chem. Phys. 2013, 138, 184102.







27 **RMG:** github.com/ReactionMechanismGenerator/RMG-Py

TCKDB: github.com/TCKDB/TCKDB

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Three types of oxidative stress testing are commonly used:

1. Using hydrogen peroxide to cover nucleophilic-electrophilic oxidations,

- 2. Using azoalkanes to cover free-radical oxidation (autoxidation), and
- 3. Using transition metals to cover single electron transfer oxidations.

AIBN (azobisisobutyronitrile)





The model was automatically generated and refined using:







Initial conditions:

5.0 mM AIBN, 12.05 mg/L N₂ (const.), 7.45 mg/L O₂ (const.)

in a water/methanol co-solvent system of 50/50 (v/v)

at 40° C, 1 atm

The model consisted of 27 species and 94 elementary reactions.

Parameters were refined using DLPNO-CCSD(T)/Def2-TZVP//@B97X-D/Def2-TZVP with solvation corrections

Major reaction pathways at 24 hours:



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35 A. Grinberg Dana et al., *Molecular Pharmaceutics* **2021**, 18(8), 3037-3049.

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36 A. Grinberg Dana et al., *Molecular Pharmaceutics* **2021**, 18(8), 3037-3049.


37 A. Grinberg Dana et al., *Molecular Pharmaceutics* **2021**, 18(8), 3037-3049.

The Radical "Soup" in API Stress Testing

Species concentrations at 72 hours:



38 A. Grinberg Dana et al., *Molecular Pharmaceutics* **2021**, 18(8), 3037-3049.

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Imipramine

Mainly used in the treatment of depression.

Also effective in treating anxiety and panic disorder.

Initial conditions:

1.96 mM API

Open system

50/50 (v/v) methanol/0.2 M carbonate buffer

pH 11 (pKa 9.5)

at 40° C

72 hours (3 days)



Important questions



Which **sites** are more susceptible to radical oxidation?

What are the major **degradation pathways** and their kinetics?

What are the chemical structures of stable degradants?

What are the relative **concentrations** of major degradants?



44 H. Wu, A. Grinberg Dana et al., *Molecular Pharmaceutics* **2022**, 19(5), 1526-1539.







•o-o

d2















46 H. Wu, A. Grinberg Dana et al., *Molecular Pharmaceutics* **2022**, 19(5), 1526-1539. **e3**

p3



47 H. Wu, A. Grinberg Dana et al., *Molecular Pharmaceutics* **2022**, 19(5), 1526-1539.



48 H. Wu, A. Grinberg Dana et al., *Molecular Pharmaceutics* **2022**, 19(5), 1526-1539.

Case Study 2: Adrenaline Thermolysis

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Initial conditions:

 $10 \,\mu\text{g/mL}$ API

Closed system

10/90 (v/v) acetonitrile / aqueous buffer

pH1 (adrenaline shown to be most stable at pH 2.5 – 4.5)

at $80^{\circ} \, \text{C}$

800 hours (~33 days)

50 Experimental data from: A. Blasko et al., *Biochemistry* 2018, 57, 4536-4546. (Novartis)

Case Study 2: Adrenaline Thermolysis





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55 H. Wu, A. Grinberg Dana et al., *Molecular Pharmaceutics* **2022**, 19(5), 1526-1539.

What can we do with predictive chemical kinetic models?



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 Assist experimental stress testing in standard preparations and result corroboration

2. Compare the chemical stability of candidate API molecules with a similar biological function





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2. Compare the chemical stability of candidate API molecules with a similar biological function

3. Screen many API molecules using a low level of theory



What can we do with predictive chemical kinetic models?

- **1.** Assist experimental stress testing in **standard preparations** and **result corroboration**
- 2. Compare the chemical stability of candidate API molecules with a similar biological function
- **3.** Screen many API molecules using a low level of theory



Remaining Challenges

- Improve the **prediction power** of the tools (i.e., add and train reaction "families")
- Improve automated lowest-energy **3D conformer** identification considering solvation effects
- Improve automated quantum-chemical computation schemes
- Add important features: chirality preservation, pH-dependent k(T), predict pKa(s) and consider protonation states, automate solvation correction schemes (COSMO-RS + Turbomol)
- Develop a high-level module to automate the entire process ("one ring to rule them all")
- Apply the self-improving tool for new APIs with various chemical motifs
- Validate against generated experimental data (seeking collaborations)

Summary

- Developed an **automated scheme** to predict API degradation
- Showed what the prominent radicals in the "Soup" system are, and their pH dependency
- Presented the first detailed kinetic model of an API degradation
- Successfully modeled the degradation of selected **APIs** (imipramine, adrenaline)
- We're developing an ecosystem of software tools and a novel platform-technology

Looking for partners capable of performing experimental API degradation studies who would like to have a front-row seat in developing, directing, and supporting a novel technological tool.



TECHNION

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Figure from AIDS Healthcare Foundation



Predictive Chemical Kinetic Modeling





ARC supports a variety of

electronic structure software





Molpro Quantum Chemistry Software







What ARC Does – Servers



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Alongside the technical challenges, the most significant

scientific challenge it to attain a reasonable TS geometry guess



KinBot: R. Van de Vijver, J. Zádor, *Comp. Phys. Comm.* 2020, 248, 106947
AutoTST: N.D. Harms, C.E. Underkoffler, R.H. West, *ChemRxiv* 2020, 10.26434/chemrxiv.13277870.v2
GCN: L. Pattanaik, J.B. Ingraham, C.A. Grambow, W.H. Green, *Phys. Chem. Chem. Phys.* 2020, 22, 2361823626
ARC: A. Grinberg Dana et al., In Preparation

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KinBot supports >30 unimolecular reaction families

KinBot: R. Van de Vijver, J. Zádor, *Comp. Phys. Comm.* 2020, 248, 106947
AutoTST: N.D. Harms, C.E. Underkoffler, R.H. West, *ChemRxiv* 2020, 10.26434/chemrxiv.13277870.v2
GCN: L. Pattanaik, J.B. Ingraham, C.A. Grambow, W.H. Green, *Phys. Chem. Chem. Phys.* 2020, 22, 2361823626
ARC: A. Grinberg Dana et al., In Preparation
GSM: P. Zimmerman, J. Chem. Theory Comput. 2013, 9(7), 3043-3050

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AutoTST supports the hydrogen abstraction, disproportionation and intra-H migration reaction families

KinBot: R. Van de Vijver, J. Zádor, Comp. Phys. Comm. 2020, 248, 106947
AutoTST: N.D. Harms, C.E. Underkoffler, R.H. West, ChemRxiv 2020, 10.26434/chemrxiv.13277870.v2
GCN: L. Pattanaik, J.B. Ingraham, C.A. Grambow, W.H. Green, Phys. Chem. Chem. Phys. 2020, 22, 2361823626
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GCN (graph convolutional network) supports isomerization reactions

KinBot: R. Van de Vijver, J. Zádor, *Comp. Phys. Comm.* 2020, 248, 106947
AutoTST: N.D. Harms, C.E. Underkoffler, R.H. West, *ChemRxiv* 2020, 10.26434/chemrxiv.13277870.v2
GCN: L. Pattanaik, J.B. Ingraham, C.A. Grambow, W.H. Green, *Phys. Chem. Chem. Phys.* 2020, 22, 2361823626
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ARC's heuristics currently only support hydrogen abstraction reactions

KinBot: R. Van de Vijver, J. Zádor, Comp. Phys. Comm. 2020, 248, 106947
AutoTST: N.D. Harms, C.E. Underkoffler, R.H. West, ChemRxiv 2020, 10.26434/chemrxiv.13277870.v2
GCN: L. Pattanaik, J.B. Ingraham, C.A. Grambow, W.H. Green, Phys. Chem. Chem. Phys. 2020, 22, 2361823626
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GSM (growing string method) is applicable for all reaction families

KinBot: R. Van de Vijver, J. Zádor, *Comp. Phys. Comm.* 2020, 248, 106947
AutoTST: N.D. Harms, C.E. Underkoffler, R.H. West, *ChemRxiv* 2020, 10.26434/chemrxiv.13277870.v2
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Additional TS Search "Adapters" are constantly being added to ARC

KinBot: R. Van de Vijver, J. Zádor, *Comp. Phys. Comm.* 2020, 248, 106947
AutoTST: N.D. Harms, C.E. Underkoffler, R.H. West, *ChemRxiv* 2020, 10.26434/chemrxiv.13277870.v2
GCN: L. Pattanaik, J.B. Ingraham, C.A. Grambow, W.H. Green, *Phys. Chem. Chem. Phys.* 2020, 22, 2361823626
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What ARC Does – TS Searches

ARC generates a TS comparison figure for each reaction, which is often insightful



* TS guesses are presented after optimization at the ω b97xd/def2svp level of theory











Scramjet Propulsion





Scramjet Propulsion

