Intro	Equivalence Classes	Correctness	Results	Markov Chains	Summary

Reduction of Graph-Grammar Models by grouping of tautomers

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Bled, February 2019



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In this presentation, we will be talking about:

- Sugar chemistry, in particular, the formose chemistry
- Equivalence Classes
- Auto catalysis
- Gillespie's exact stochastic simulation algorithm
- Bisimulation
- Markov Chains



- Derivation graphs can get LARGE
- Many nodes may have extremely similar properties
- Stack these nodes together in a projection

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Why d	o this?				

- Projected derivation graphs have much fewer nodes
- Much faster auto catalysis computation, and more correct results
- Stochastic simulations run significantly faster, and potentially produce more correct results









Formos	e reaction	autocatalysis	example:	Projection	
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- $\{x \in X : xRy\}$
- y is an element of X
- The notation "*xRy*" means there is an equivalence relation between *x* and *y*
- for all $x, y \in X$, we have $xRy \Leftrightarrow x$ and y belong to the same equivalence class

M/hy	is projected a	ito catalvei	s more co	rrect?	
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oWhy could projected stochastic simulation be more
correct?

- Keto-Enol reactions are said to be extremely fast, especially comapred to Aldol-addition reactions
- Getting a realistic relative reaction rate, and also a reasonable computation time, is not possible normally
- "Removing" the Keto-Enol reactions from the equation makes it much easier to get something closer to real life

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Bisimilarity definitions:

- Strong Bisimilarity: A binary relation R over the set of states of an Labelled Transition System is a strong bisimulation if and only if whenever s₁Rs₂ and α is an action:
 - if $s_1 \xrightarrow{\alpha} s_1'$, then there is a transition $s_2 \xrightarrow{\alpha} s_2'$ such that $s_1' \mathcal{R} s_2'$, and
 - if $s_2 \xrightarrow{\alpha} s_2'$, then there is a transition $s_1 \xrightarrow{\alpha} s_1'$ such that $s_1' \mathcal{R} s_2'$
- Weak Bisimilarity: A binary relation \mathcal{R} over the set of states of a Labelled Transition System is a weak bisimulation if and only if whenever $s_1\mathcal{R}s_2$ and α is an action (Including τ , the internal action):
 - If $s_1 \xrightarrow{\alpha} s_1'$, then there is a transition $s_2 \xrightarrow{\alpha} s_2'$ such that $s_1' \mathcal{R} s_2'$ and
 - If $s_2 \xrightarrow{\alpha} s'_2$, then there is a transition $s_1 \stackrel{\alpha}{\Rightarrow} s'_1$ such that $s'_1 \mathcal{R}s'_2$.

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Concur	rency Theory:	Bisimula	tions, exai	mple	



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- A "Unique reaction" in the following table means specifically a 1-2 or 2-1 reaction.
- The autocatalysis results are best described by how many different solutions are found and how fast. This is reflected in how much of the table is actually filled out.
- The stochastic simulation results are best described by how fast a simulation reaches its end, and how far the simulation got towards its end.

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Cells: How many different solutions

Columns: How many carbons are contained in the largest molecule in the solution

	4 C	5 C	6 C	7 C	8 C	9 C	Sum
3 Unique Reactions	1	0	0	0	0	0	1
4 Unique Reactions	0	2	8	4	6	12	32
5 Unique Reactions	0	32	119	394	927	—	1472
6 Unique Reactions	1	36	412	1640	4880	—	6969
7 Unique Reactions	0	92	2556	25586	—	—	28234
8 Unique Reactions	0	144	10053	137364	—	—	147561
9 Unique Reactions	0	185	45469			—	45654
10 Unique Reactions	0	239	170536	—		—	170775

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	4 C	5 C	6 C	7 C	8 C	9 C	Sum
3 Unique Reactions	1	0	0	0	0	0	1
4 Unique Reactions	0	0	2	0	0	0	2
5 Unique Reactions	0	18	29	106	196	357	706
6 Unique Reactions	1	21	162	561	1278	—	2023
7 Unique Reactions	0	57	891	7271	27768	—	35987
8 Unique Reactions	0	102	4012	45817		—	49931
9 Unique Reactions	0	125	17529	377398			395052
10 Unique Reactions	0	166	67407	—		—	67573

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 Stochastic simulation results:
 Unprojected

Runtime: 81 minutes and 51.395 seconds Details: 100 times higher reaction rate on keto-enol than on aldol-addition





Runtime: 6 minutes and 26.545 seconds

Details: Same reaction rate on aldol-addition as for unprojected





Runtime: 25 minutes and 18.185 seconds Details: Keto-enol reaction rate same as aldol-addition





Runtime: 3 minutes and 42.391 seconds Details: reaction rate 1/5th of unprojected.



Markov	Chains.	Example 1.	Formose rea	action	
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Markov	Chains:	Example 1:	Formose	reaction	





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Summa	ary				

- Equivalence classes can be used to reduce derivation graphs based on (for instance) tautomerisms
- Such projections improve both the computation time and quality of auto catalysis and
- also improves the computation time of Stochastic simulations, and may sometimes also improve the quality of the stochastic simulation itself
- Markov Chains can be used to show that the projection does not appear to cause any negative effect on the concentration differences in such a stochastic simulation
- The comparison between the unprojected and projected solutions can be proven to be weakly bisimilar