



UNIVERSITY OF
BIRMINGHAM

ENTERPRISE

Omics-based Chemical Grouping

- ◉ Professor Mark Viant
- ◉ Michabo Health Science - University of Birmingham Enterprise
- ◉ Michabo Health Science Ltd

- ◉ BTS/NC3R/HSE CRD virtual workshop on NAMs
- ◉ 23 February 2022

Research team



Rosemary Barnett
Tom Lawson
Elena Sostare
John Colbourne



Doris Hirmann
Tomasz Sobanski
Bram Versonnen
Mounir Bouhifd



Hanna Gruszczynska
Jiarui Zhou



Ralf Weber
Gavin Lloyd

Disclosure: Mark Viant is a co-founder of Michabo Health Science Ltd

mark@michabo.co.uk

Outline

Rationale for case study

Study objectives

Results

- Conventional vs. omics-based grouping
- Read-across
- Predicting the structural drivers of omics-based grouping

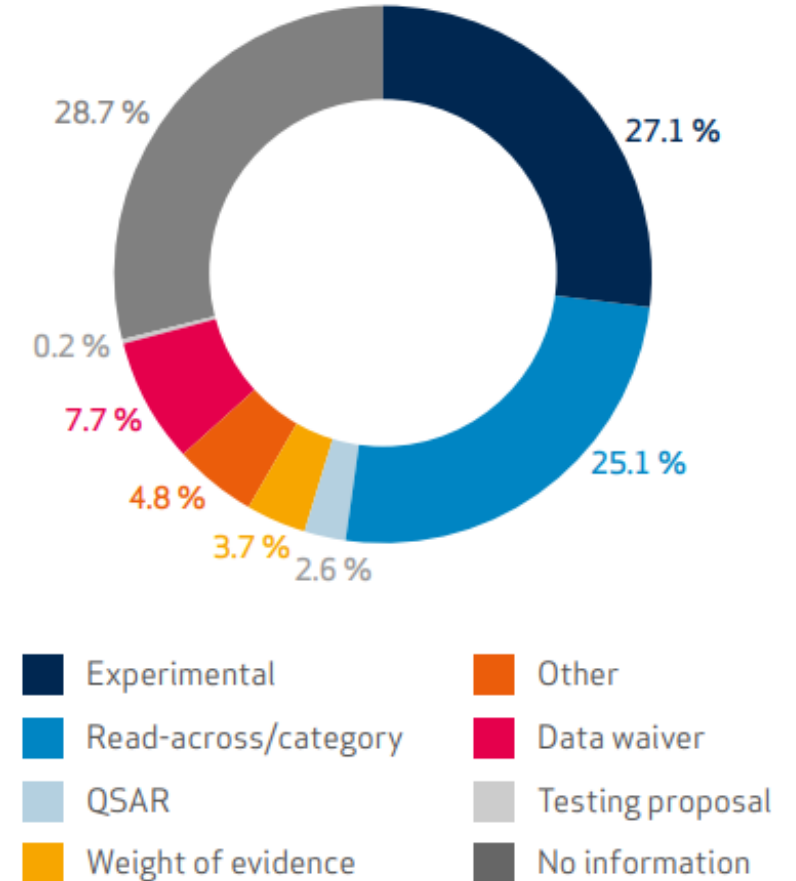
Conclusions from case study

Challenges and where next?



Grouping and read-across

- **Most commonly used** alternative method - for data-gap filling (https://echa.europa.eu/documents/10162/0/alternatives_test_animals_2020_en.pdf)
- Reduces the need for experimental tests because information on a similar substance (**source**) is used to predict the properties of another substance (**target**)
- When Test Guideline studies have been used to generate the data for the source substance, then a properly justified read-across can be used to **fulfil REACH information requirements**
- While ECHA has advocated using grouping/read-across, it has had to **reject the majority of read-across arguments** due in part to lack of scientific rigour in defining groups of substances, leading to incompliance



Can we increase the scientific evidence and therefore the acceptance rate of grouping/read-across dossiers by substantiating them with grouping based upon molecular mechanistic data?



View presentation from one of 2 perspectives

1. NAM (omics)-enhanced grouping **to enable read-across**
2. **Omics-based grouping** to support the acceleration of chemical risk assessment, first using 'omics data to screen (e.g. *in vitro*) and then group a large number of substances based on their MoA, prioritising group-representative substances for higher tier testing

'Group first...' – H2020 PrecisionTox

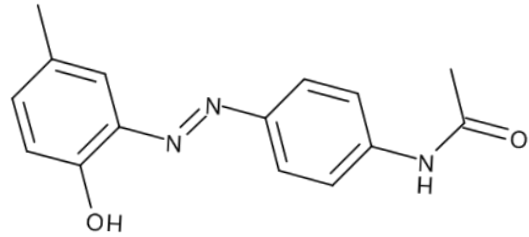


Study objectives

1. Select a target substance and series of potential source substances for grouping/read-across (analogue approach)
2. Apply **conventional approaches** to form a grouping hypothesis
3. Apply **omics approaches** to substantiate or disprove this grouping hypothesis based on **molecular mechanistic data** (here using transcriptomics and metabolomics)
4. Conduct read-across to fill the data gap
5. Map ToxPrint chemotypes onto omics-based grouping to predict what structural features are driving that biologically-based grouping

Azo dyes

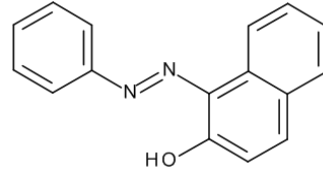
Target substance:



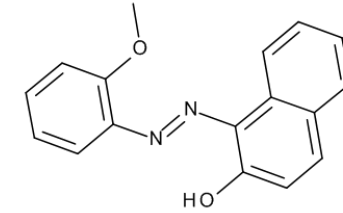
Disperse yellow 3 (DY3)



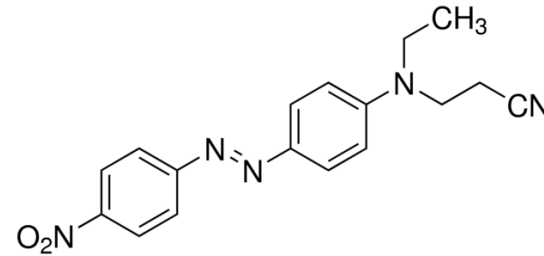
Six potential source substances:



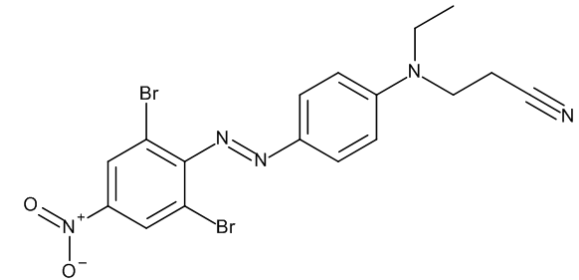
Sudan 1 (S1)



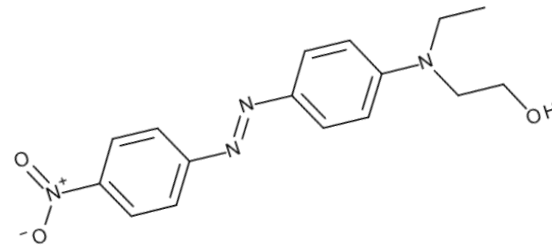
Sudan red G (SRG)



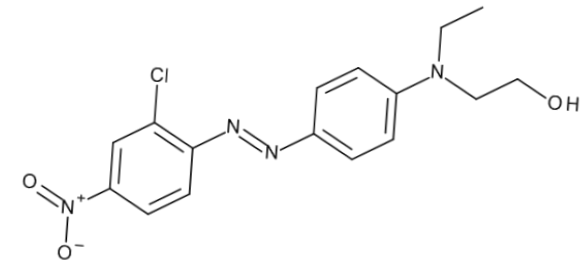
Disperse orange 25 (DO25)



Disperse orange 61 (DO61)



Disperse red 1 (DR1)



Disperse red 13 (DR13)

Need to determine which source substance is most similar to the target

Grouping / read-across scenario

Type: ECHA's Read-Across Assessment Framework, Scenario 2 - analogue approach with single target substance and single source substance

Test system:



Daphnia magna

Expt'al design: low, medium, high doses for each azo dye

Endpoint (to read-across): *Daphnia* chronic reproductive toxicity (OECD TG211)

Principal aim of study was to investigate omics-based grouping, irrespective of the biological test system and test substances

Conventional approaches to form grouping hypothesis (1)

QSAR TOOLBOX

The OECD QSAR Toolbox
for Grouping Chemicals
into Categories

- OASIS, ECOSAR 2.0, EPA and OECD chemical categories

DY3	Reactive unspecified alert by acute aquatic toxicity MOA (OASIS) Belongs to Phenols , Amides, Phenol amines (ECOSAR 2.0) Belongs to Phenols (EPA New Chemical Categories) Belongs to m,p-Cresols (OECD HPV Chemical Categories)
S1, SRG	Reactive unspecified alert by acute aquatic toxicity MOA (OASIS) Belong to Phenols (ECOSAR 2.0)
DR1, DR13, DO25, DO61	Reactive unspecified alert by acute aquatic toxicity MOA (OASIS) Belong to Neutral organics (ECOSAR 2.0) Only DR13 and DO61 belong to Neutral organics (EPA New Chemical Categories)

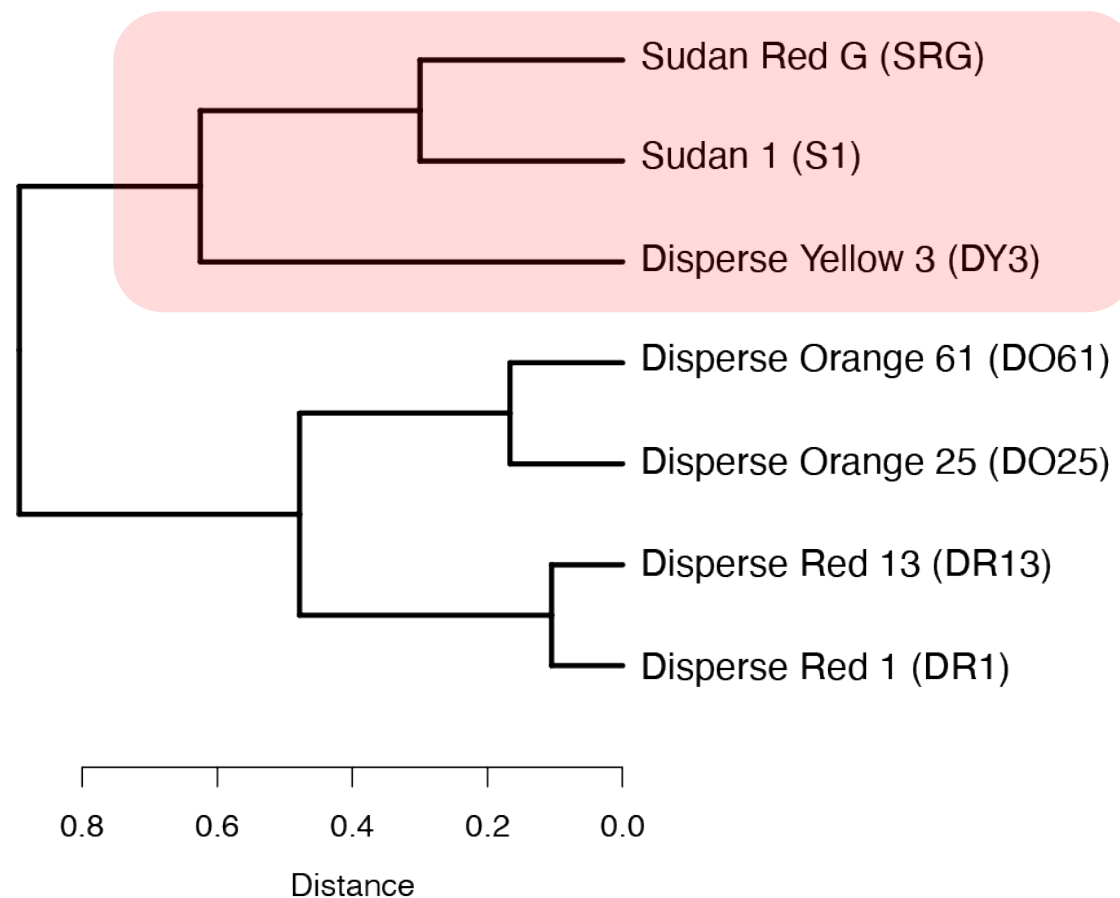
- **Conclude: DY3 lies in its own group (but note DY3, S1 and SRG are all 'phenols')**

'Conventional' approaches to form grouping hypothesis (2)

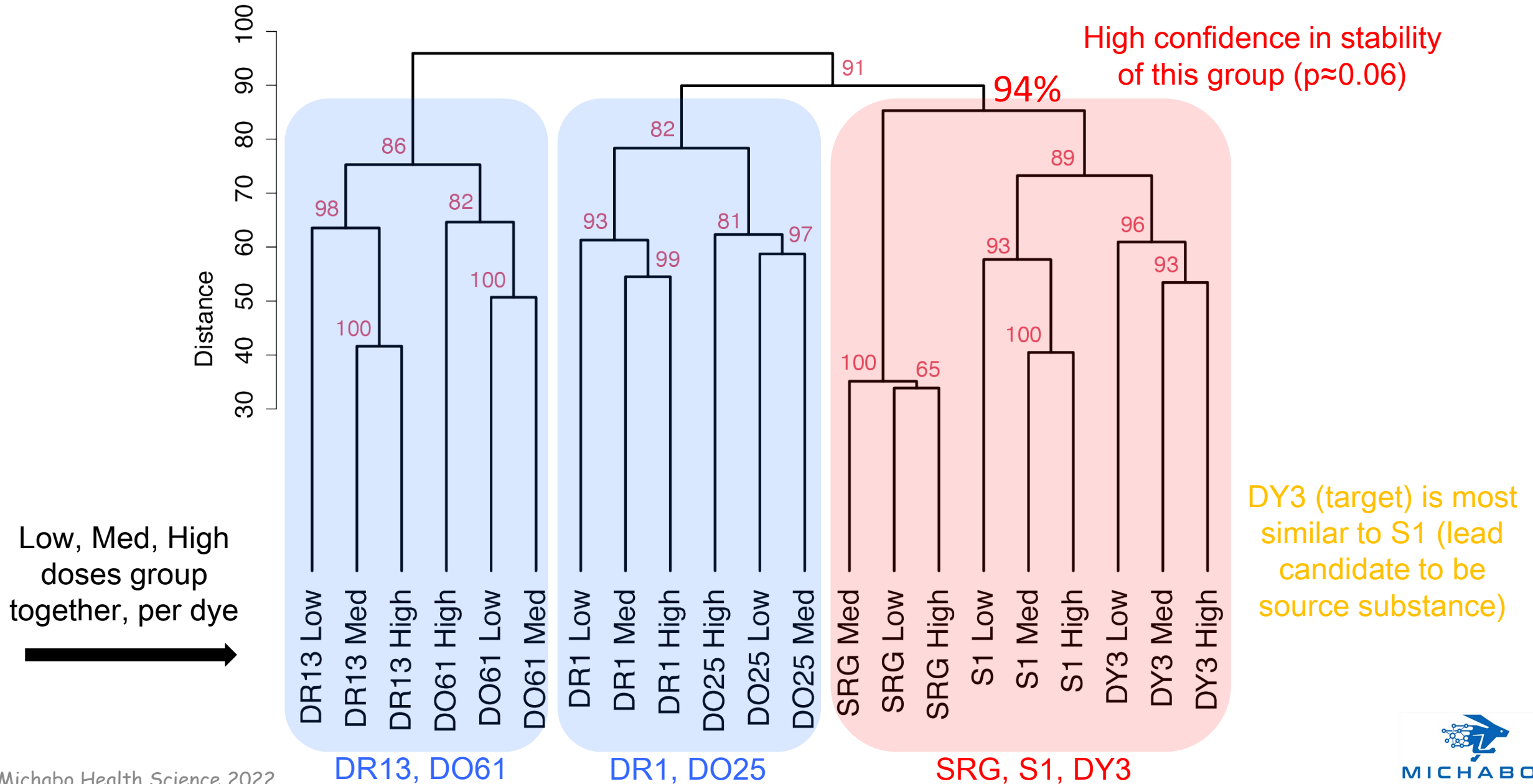
ToxPrint chemotypes

- Structural fingerprints containing 729 binary descriptors (atoms, bonds, rings, functional groups)
- Tanimoto distance matrix
- Hierarchical cluster analysis

Conclude: consistent with QSAR profiling, DY3 is quantitatively more similar to S1 and SRG than to DR1, DR13, DO25 and DO61



Multi-omics approach to form grouping hypothesis



Mechanistic anchoring of omics data - transcriptomics



“Cellular stress and injury”
pathway group



- GP6 Signaling Pathway
- EIF2 Signaling
- RhoA Signaling
- TNFR1 Signaling
- Antiproliferative Role of TOB in T Cell Signaling
- Oxidative Phosphorylation
- Role of RIG1-like Receptors in Antiviral Innate Immunity
- Natural Killer Cell Signaling
- Neuroinflammation Signaling Pathway
- Regulation Of The Epithelial Mesenchymal Transition In Development Pathway
- NRF2-mediated Oxidative Stress Response
- IL-1 Signaling
- CD28 Signaling in T Helper Cells
- NF-κB Activation by Viruses
- iCOS-iCOSL Signaling in T Helper Cells
- Type II Diabetes Mellitus Signaling
- FAT10 Cancer Signaling Pathway
- MIF Regulation of Innate Immunity
- Xenobiotic Metabolism AHR Signaling Pathway
- Acute Phase Response Signaling
- 4-1BB Signaling in T Lymphocytes
- Role of NFAT in Regulation of the Immune Response
- TNFR2 Signaling
- Type I Diabetes Mellitus Signaling
- Breast Cancer Regulation by Stathmin1
- Coronavirus Pathogenesis Pathway

SRG_H
DY3_H
S1_H



SRG, S1 and DY3
induce similar pathway
perturbations

Mechanistic anchoring of omics data - metabolomics






SOT | Society of
Toxicology
academic.oup.com/toxsci

TOXICOLOGICAL SCIENCES, 2022, 1–13

<https://doi.org/10.1093/toxsci/kfac007>
Advance Access Publication Date: 30 January 2022
Research article

Knowledge-Driven Approaches to Create the MTox700+ Metabolite Panel for Predicting Toxicity

Elena Sostare ^{*}, Thomas N. Lawson ^{*}, Lucy R. Saunders,[†] John K. Colbourne,^{*,†} Ralf J. M. Weber,[†] Tomasz Sobanski,[‡] and Mark R. Viant ^{*,†,1}

^{*}Michabo Health Science, University of Birmingham Enterprise, Birmingham Research Park, Birmingham B15 2SQ, UK, [†]School of Biosciences, University of Birmingham, Birmingham B15 2TT, UK, and [‡]European Chemicals Agency (ECHA), 00150 Helsinki, Finland

¹To whom correspondence should be addressed at Michabo Health Science, University of Birmingham Enterprise, Birmingham Research Park, Vincent Drive, Birmingham B15 2SQ, UK. E-mail: mark@michabo.co.uk

Grouping hypothesis based on phys-chem and 'omics data

	Grouping/read-across workflow	Finding
1	Conventional grouping <ul style="list-style-type: none">• QSAR profilers• Clustering ToxPrint chemotypes	DY3 might be in its own group, but is most similar to S1 and SRG
2	Omics-based grouping <ul style="list-style-type: none">• Clustering multi-omics profiles• Molecular pathway perturbations	DY3 in a group with S1 and SRG, and most similar to S1
3	Final grouping hypothesis	Source = S1 Target = DY3

Read-across to predict toxicity

Sudan 1 (S1)



Disperse yellow 3 (DY3)



Endpoint: Daphnia chronic reproductive toxicity (OECD TG211)

NOEC (measured) $\approx 40 \mu\text{g/L}$

LOEC (measured) $\approx 60 \mu\text{g/L}$



Fill data gap in the hazard characterisation of DY3

NOEC (predicted) $\approx 40 \mu\text{g/L}$

Experimentally confirmed the prediction:

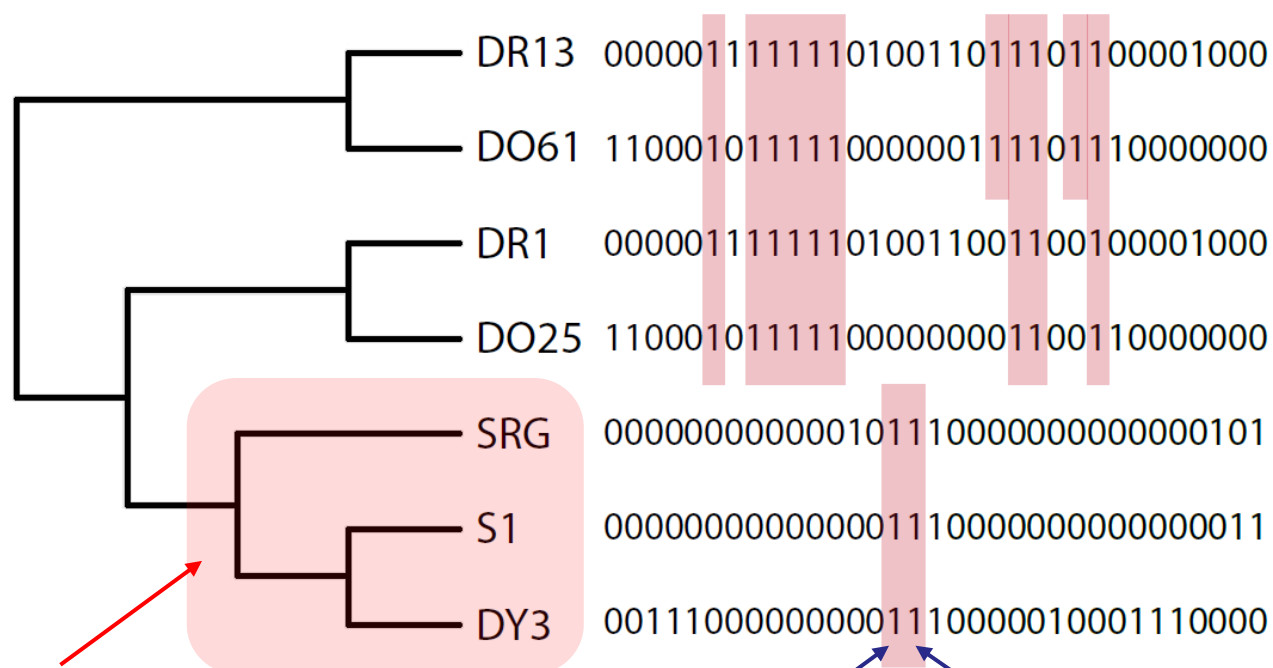
NOEC (*measured*) $\approx 40 \mu\text{g/L}$

LOEC (*measured*) $\approx 75 \mu\text{g/L}$

Map ToxPrint chemotypes onto Omics-based grouping

Dendrogram derived from multi-omics grouping

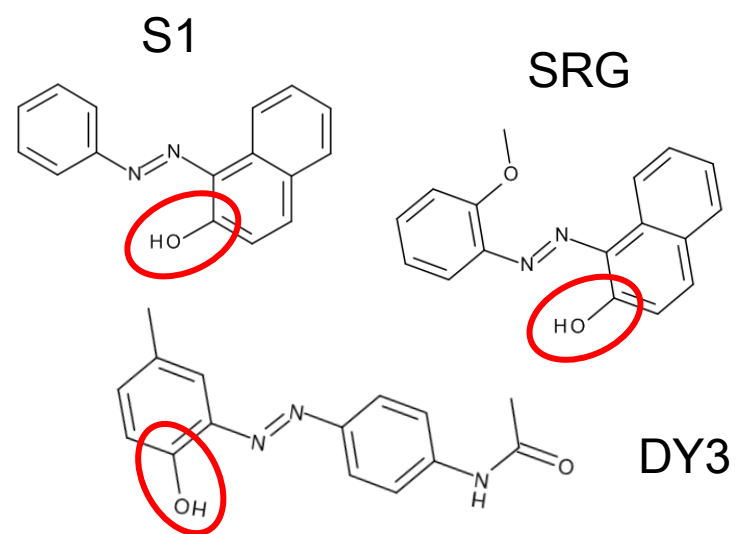
ToxPrint chemotypes – showing only the non-zero structural fragments (33 of 729)



Only 2 structural fragments are unique to this branch

bond: COH_alcohol_aromatic

bond: COH_alcohol_aromatic_phenol



Conclusions from Case Study

- Established a workflow that enables NAM ('omics) mechanistic data to be used alongside conventional grouping approaches
- Demonstrated how 'omics data can provide a quantitative measure of similarity, allowing 7 azo dyes to be reliably grouped, and an optimal source substance identified for read-across
- Experimentally confirmed the read-across prediction
- Using ToxPrint chemotypes, predicted that aromatic phenols are driving this biologically-based grouping
- Paper in preparation

Where next? – Challenges for omics-based grouping

- **Mechanistic anchoring** of molecular responses to MoA
- **Acceptability (“validation”)** of ‘omics applications
 - Reproducibility / reliability
 - Cefic MATCHING international ring-trial (metabolomics, chemical grouping) - *ongoing*
 - Tiered criteria based on Context of Use (CoU) of metabolomics - *ongoing*
- **Reporting** of ‘omics
 - OECD Omics Reporting Framework (TRF, MRF) - DOI: [10.1016/j.yrtph.2021.105020](https://doi.org/10.1016/j.yrtph.2021.105020)
 - OECD WPHA proposal *under review* - ‘omics-based chemical grouping
- **Clear, extensive documentation** of omics-based grouping
 - Horizon 2020 PrecisionTox task