Machine-Learning Mathematical Structures

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M-Theory and Mathematics: classical and quantum aspects NYU Abu Dhabi, Jan 2023

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Enriching the Maths/Physics Dialogue

- Alg./diff. Geometry/topology Rep. Theo : the right language for physics
 - Gravity ~ Ricci 2-form of Tangent bundles;
 - Elementary Particles ~ irred reps of the Lorentz group and sections of bundles with Lie structure group; Interactions ~ Tensor products of sections ...
 - String theory: brain-child of gauge-gravity geometrization tradition
- A new exciting era for synergy with (pure & computational) geometry, group theory, combinatorics, number theory: *Sage*, *M2*, *GAP*, *LMFDB*, *GrDB* are becoming indispensible tools for physicists
- Interdisciplinary enterprise: cross-fertilisation of particle/string theory, phenomenology, pure mathematics, computer algorithms, data-bases, ...

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Vacuum Degeneracy

Perhaps the biggest theoretical challenge to string theory: selection criterion??? metric on the landscape???

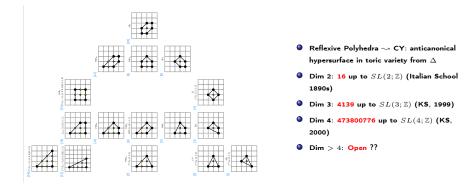
- Douglas (2003): Statistics of String vacua
- Kachru-Kallosh-Linde-Trivedi (2003): type II/CY estimates of 10^{500}
- Taylor-YN Wang (2015-7): F-theory estimates 10^{3000} to 10^{10^5}
- Basic Reason:

Algebraic Geometry \rightsquigarrow Combinatorial Geometry \rightsquigarrow Exponential Growth in dim

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e.g., Borisov-Batyrev & Kreuzer-Skarke



GrDB: Brown, Kaspryzyk, Nil, Kahle, ... http://www.grdb.co.uk/ Altman-Gray-YHH-Jejjala-Nelson (2014): brute-force: $\sim 10^6$ up to $h^{1,1} = 6$ Altman-Carifio-Halverson-Nelson (2018): estimated 10^{10^4} triangulations

Demirtas-Long-McAllister-Stillman (2019): all triang $240 \le h^{1,1} \le 491$

Image: A math a math

2017: String Theory enters the Machine-Learning Era

YHH (1706.02714); Krefl-Seong (1706.03346); Ruehle (1706.07024); Carifio-Halverson-Krioukov-Nelson (1707.00655)



Sophia: Hanson Robotics, HongKong

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- Beginning of String_Data Annual conference series
- How can ML and modern data-science help with the vacuum degeneracy problem??

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- Beginning of String_Data Annual conference series
- How can ML and modern data-science help with the vacuum degeneracy problem??
- Meanwhile ... Sophia becomes a "human" citizen (in Saudi Arabia)

Progress in String Theory: Start Dates of Annual Series

- 1986- "Strings" Conference
- 2002- "StringPheno" Conference
- 2006 2010 String Vacuum Project (NSF)
- 2008 ISGT Integrability in String/Gauge
- 2011- "String-Math" Conference (2020 , M-theory & Maths Workshop)
- 2012- "Amplitudes"
- 2014- String/Theoretical Physics Session in SIAM Conference
- 2017- "String-Data" Conference (2022: Cambridge (Organizers Berman, YHH, Heyes, Hirst, Mishra)

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Algebraic Geometry as Image Processing

• A typical calculation:

- Key to computational Algebraic Geometry: Gröbner basis, double-exponential complexity (unlike Gaussian elimination which is generalizes)
- [YHH 1706.02714] Deep-Learning the Landscape, *PLB 774, 2017*; (cf. Feature in *Science*, Aug, vol 365 issue 6452, 2019): think of it as an

image processing problem



from String Landscape to the Mathematical Landscape

Machine Learning Mathematical Structures

Why stop at string/geometry?

q.v. Review Paper: YHH 2101.06317

• $[0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, \ldots]$

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- [0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, ...] multiple of 3 or not.
- $[1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, \dots]$

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Pattern Recognition: Machine-Learning

 Binary Classification of a Binary Vector (sliding window of, say, length 100); supervised learning: predict next one, e.g., Prime/Not becomes:

$\{0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,$	\rightarrow	1
$\{1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, \dots, 1\}$	\rightarrow	0
$\{0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, \dots, 0\}$	\rightarrow	1

Image: A matrix

Pattern Recognition: Machine-Learning

 Binary Classification of a Binary Vector (sliding window of, say, length 100); supervised learning: predict next one, e.g., Prime/Not becomes:

- pass to standard classifiers: SVW, Bayes, Nearest Neighbour; NN of the form $\mathbb{R}^{100} \xrightarrow{\text{linear}} \mathbb{R}^{20} \xrightarrow{\text{tanh}} \mathbb{R}^{20} \xrightarrow{\text{Round} \sum} \mathbb{Z}$, your kitchen sink, ...
- take 50,000 samples, 20-80 cross-validation, record (precision, MCC)
- similar performance for most: Mod3: (1.0, 1.0); PrimeQ, after balancing: (0.8, 0.6); Liouville Λ: (0.5, 0.001)

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Thank you! Since 2017-

my fantastic students

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Sutherland, Eldar Sultanow

Representation Theory: Mandy Cheung, Pierre Dechant, Minhyong Kim, Jianrong Li, Gregg Musiker

Combinatorics: Johannes Hofscheier, Alexander Kasprzyk, Shiing-Tung Yau

ML Maths

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How does one *DO* mathematics, I ?

Russell-Whitehead Principia Mathematica [1910s] (Leibniz, Frege, ...) axiomatize maths, but ... Gödel [1931] Incompleteness; Church-Turing [1930s] Undecidability
 Automated Theorem Proving (ATP) "The practicing mathematician hardly ever worries about Gödel"

- Newell-Simon-Shaw [1956] Logical Theory Machine: subset of Principia
- Type Theory [1970s] Martin-Löf, Coquand, ... Coq: 4-color (2005);
 Feit-Thompson Thm (2012); Lean (2013); Univalent Foundation / Homotopy Type Theory [2006-] Voevodsky

Buzzard: "Future of Maths" 2019, ICM 2022 Davenport: ICM 2018

"Computer Assisted Proofs" Szegedy: more extreme view, computers > humans @ chess (1990s); @ Go (2018); @ Proving theorems (2030)

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We can call this Bottom-up Mathematics

How does one *DO* mathematics, II ?

• Historically,

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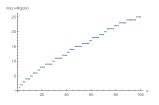
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How does one *DO* mathematics, II ?

- Historically, Maths perhaps more Top-Down: practice before foundation
 - Countless examples: calculus before analysis; algebraic geometry before Bourbaki, permutation groups / Galois theory before abstract algebra ...
 - ▶ A lot of mathematics starts with intuition, experience, and experimentation
- The best neural network of C18-19th?

How does one *DO* mathematics, II ?

- Historically, Maths perhaps more Top-Down: practice before foundation
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 - A lot of mathematics starts with intuition, experience, and experimentation
- The best neural network of C18-19th? brain of Gauß ; e.g., age 16



(w/o computer and before complex analysis [50 years before Hadamard-de la Vallée-Poussin's proof]): PNT $\pi(x) \sim x/\log(x)$

• BSD computer experiment of Birch & Swinnerton-Dyer [1960's] on plots of rank r & N_p on elliptic curves

Example I: Representation/Group Theory

- ML Algebraic Structures (GAP DB) [YHH-MH. Kim 1905.02263,]
 - When is a Latin Square (Sudoku) the Cayley (multiplication) table of a finite group? Bypass quadrangle thm (0.95, 0.9)
 - Can one look at the Cayley table and recognize a finite simple group?
 - bypass Sylow and Noether Thm; (0.97, 0.95) rmk: can do it via character-table T, but getting T not trivial
 - SVM: space of finite-groups (point-cloud of Cayley tables) seems to exist a hypersurface separating simple/non-simple

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Example II: Combinatorics

[YHH-ST. Yau 2006.16619] Wolfram Finite simple graphs DB

ML standard graph properties:

?acyclic (0.95, 0.96); ?planar (0.8, 0.6); ?genus >, =, < 0 (0.8, 0.7); ?∃
Hamilton cycles (0.8, 0.6); ?∃ Euler cycles (0.8, 0.6)
(Rmk: NB. Only "solving" the likes of traveling salesman stochastically)</pre>

- spectral bounds $(R^2 \sim 0.9) \dots$
- Recognition of Ricci-Flatness (0.9, 0.9) (todo: find new Ricci-flat graphs);

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Example III: Quivers, Clusters, Brane setups, ...

- [Bao-Franco-Hirst-Musiker, 2006.10783, Dechant-YHH-Heyes-Hirst 2203.13847] Recognition of mutation types (> 0.9)
- [Hirst-YHH-Peterken 2004.05218]: adjacency+permutation triple of dessin d'enfants; predicting transcendental degree > 0.9
- [Arias-Tamargo, YHH, Heyes, Hirst, Rodriguez-Gomez 2202.05845] Recognition of equivalence (SL(2; ℤ), Seiberg, Hanany-Witten) of brane-webs
- [Cheung-Dechant-YHH-Heyes-Hirst-Li 2212.09771] learning Young tableaux representation of variables in Grassmannian cluster algebras (> 0.99)

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Example IV: Number Theory

Arithmetic, A Classical Reprobate?

- [YHH 1706.02714, 1812.02893:]
 - Predicting primes $2 \rightarrow 3, 2, 3 \rightarrow 5, 2, 3, 5 \rightarrow 7$; no way
 - PrimeQ: (0.7, 0.8); Sarnak's Challenger of Liouville Lambda (0.5, 0.001)
- [Alessandretti-Baronchelli-YHH 1911.02008]

ML/TDA@Birch-Swinnerton-Dyer III and Ω ok with regression & decision trees: RMS < 0.1; Weierstrass \rightarrow rank: random

- Arithmetic Geometry: A Modern Hope? YHH-KH Lee-Oliver
 - ▶ 2010.01213: Complex Multiplication, Sato-Tate (0.99 ~ 1.0, 0.99 ~ 1.0)
 - 2011.08958: Number Fields: rank and Galois group (0.97, 0.9)
 - 2012.04084: BSD from Euler coeffs, integer points, torsion (0.99, 0.9); Tate-Shafarevich III (0.6, 0.8) [Hardest quantity of BSD]

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Clearly useful for maths and physics

looking for new conjectures e.g.,

- '19 YHH-Kim: separating hyperplane simple/non-simple groups; open
- '19 Brodie-Constantin-Lukas: exact formulae for cohomo surf.; proved.
- '20 YHH-Lee-Oliver: L-coefs and integer pt./torsion on ell; Known.
- '20 Craven-Jejjala-Par: Jones poly best-fit function; open
- '22 DeepMind-Oxford-Sydney, Nature: Volume bounds for knots; proved

speed up computations and accuracies e.g.,

- computing/estimating (top.inv., charges, etc) MUCH FASTER
- '19 Ashmore-YHH-Ovrut: speed up Donaldson alg@CY metric 10-100
- '20 Douglas et al., Anderson et al. improves Donaldson 10-100 times

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An Inherent Hierarchy?

• In decreasing precision/increasing difficulty:

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\begin{array}{rl} \mbox{numerical} \\ \mbox{string theory} \rightarrow & \mbox{algebraic geometry over } \mathbb{C} \sim \mbox{arithmetic geometry} \\ & \mbox{algebra} \\ \mbox{string theory} \rightarrow & \mbox{combinatorics} \\ & \mbox{analytic number theory} \end{array}
```

Image: A matrix

Please submit

Launching in 2023

IJDSMS

Calling for Papers

Editor-in-Chief Yang-Hui He London Institute for Mathematical Sciences & Merton College, University of Oxford email: hey@maths.ox.ac.uk

More Information: https://www.worldscientific.com/worldscinet/ijdsms



INTERNATIONAL JOURNAL OF DATA SCIENCE IN THE MATHEMATICAL SCIENCES

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Meta-mathematics/physics?

[YHH-Jejjala-Nelson] "hep-th" 1807.00735

Word2Vec: [Mikolov et al., '13] NN which maps words in sentences to a vector space by context (much better than word-frequency, quickly adopted by Google); maximize (partition function) over all words with sliding window (W_{1,2} weights of 2 layers, C_α window size, D # windows)

$$Z(W_1, W_2) := \frac{1}{|D|} \sum_{\alpha=1}^{|D|} \log \prod_{c=1}^{C_{\alpha}} \frac{\exp([\vec{x}_c]^T \cdot W_1 \cdot W_2)}{\sum_{j=1}^{V} \exp([\vec{x}_c]^T \cdot W_1 \cdot W_2)}$$

We downloaded all ~ 10⁶ titles of hep-th, hep-ph, gr-qc, math-ph, hep-lat from ArXiv since the beginning (1989) till end of 2017 (word cloud (rmk: Ginzparg has been doing a version of linguistic ML on ArXiv) (rmk: abs and full texts in future)
 YANG-HUL HE (London/Oxford/(Nankai))

Subfields on ArXiv has own linguistic particulars

• Linear Syntactical Identities

bosonic + *string-theory* = *open-string*

holography + quantum + string + ads = extremal-black-hole

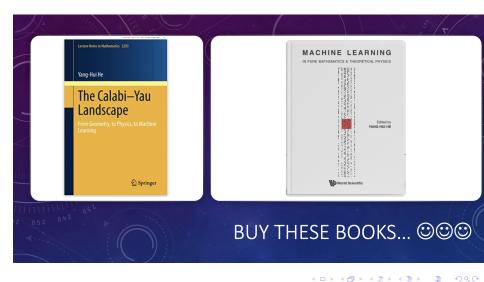
string-theory + calabi-yau = m-theory + g2

space + black-hole = geometry + gravity ...

- binary classification (Word2Vec + SVM) of formal (hep-th, math-ph, gr-qc) vs phenomenological (hep-ph, hep-lat) : 87.1% accuracy (5-fold classification 65.1% accuracy). ArXiv classifications
- Cf. **Tshitoyan et al.**, "Unsupervised word embeddings capture latent knowledge from materials science literature", **Nature** July, 2019: 3.3. million materials-science abstracts; uncovers structure of periodic table, predicts discoveries of new

thermoelectric materials years in advance, and suggests as-yet unknown materials < \ge > > \ge < > < <

Please buy



The London Institute for Mathematical Sciences

- UK's only independent research institute for maths; modelled after IAS, Princeton
- Founded in 2011 by Dr. Thomas Fink
- Housed in the Faraday Suites of the Royal Institution of Great Britain
- 1 of 23 themes: AI for Maths Discovery

https://lims.ac.uk/event/ai-assisted-maths-discovery/

• Just established:

Arnold Felowships Landau Fellowships









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ML Math

ML Maths, Jan, 2023 24 / 30

THANK YOU!

YANG-HUI HE (London/Oxford/Nankai)

3

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The Proper Way $\mathcal{O}(e^{e^d})$

• Recall Hodge decomposition $H^{p,q}(X) \simeq H^q(X, \wedge^p T^\star X) \rightsquigarrow$

 $H^{1,1}(X) = H^1(X, T_X^*), \qquad H^{2,1}(X) \simeq H^{1,2} = H^2(X, T_X^*) \simeq H^1(X, T_X)$

• Euler Sequence for subvariety $X \subset A$ is short exact:

$$0 \to T_X \to T_M|_X \to N_X \to 0$$

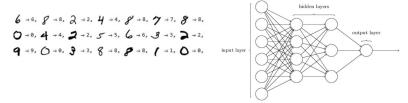
Induces long exact sequence in cohomology:

• Need to compute Rk(d), cohomology and $H^i(X, T_A|_X)$ (Cf. Hübsch)

Back to Alg Geo YANG-HUI HE (London/Oxford/Nankai)

The Neural Network Approach

- Bijection from 1234567890 to $\{1, 2, \dots, 9, 0\}$?
- Take large sample, take a few hundred thousand (e.g. NIST database)



• Data = Training Data ⊔ Validation Data

Test trained NN on validations data to see accuracy performance

Universal Approximation Theorems

Large Depth Thm: (Cybenko-Hornik) For every continuous function $f : \mathbb{R}^d \to \mathbb{R}^D$, every compact subset $K \subset \mathbb{R}^d$, and every $\epsilon > 0$, there exists a continuous function $f_\epsilon : \mathbb{R}^d \to \mathbb{R}^D$ such that $f_\epsilon = W_2(\sigma(W_1))$, where σ is a fixed continuous function, $W_{1,2}$ affine transformations and composition appropriately defined, so that $\sup_{x \in K} |f(x) - f_\epsilon(x)| < \epsilon$.

- Large Width Thm: (Kidger-Lyons) Consider a feed-forward NN with n input neurons, m output neuron and an arbitrary number of hidden layers each with n + m + 2 neurons, such that every hidden neuron has activation function φ and every output neuron has activation function the identity. Then, given any vector-valued function f from a compact subset $K \subset \mathbb{R}^m$, and any $\epsilon > 0$, one can find an F, a NN of the above type, so that $|F(x) f(x)| < \epsilon$ for all $x \in K$.
- **ReLU Thm:** (Hanin) For any Lebesgue-integral function $f : \mathbb{R}^n \to \mathbb{R}$ and any $\epsilon > 0$, there exists a fully connected ReLU NN F with width of all layers less than n + 4 such that $\int_{\mathbb{R}^n} |f(x) F(x)| dx < \epsilon$.

Back to NN@Alg Geo

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ArXiv Word-Clouds

A field a fair water and a fair a





hep-ph



math-ph

Back to Word2Vec

(日) (四) (三) (三)

gr-qc

Classifying Titles

Compare, + non-physics sections, non-science (Times), pseudo-science (viXra)

	Word2Vec + SVM			1	2	3	4	5					
Actual				_	_	Ŭ					1	:	hep-th
	1			40.2	6.5	8.7	24.0	20.	6		2	:	hep-ph
	2			7.8	65.8	12.9	9.1	4.4	1		3	:	hep-lat
	3			7.5	11.3	72.4	1.5	7.4	1		4	:	gr-qc
	4			12.4	4.4	1.0	72.1	10.	2		(5	:	math-ph
	5			10.9	2.2	4.0	7.8	75.1					
	Actual	NN	1	2	3	4	5	6	7	8	9	10	
·	viXra-	hep	11.5	5 47.4	6.8	13.	11.	4.5	0.2	0.3	2.2	3.1	
	viXra-o	qgst	13.3	3 14.5	1.5	54.	8.4	1.8	0.1	1.1	2.8	3.	
6: cond-mat, 7: q-fin, 8: stat, 9: q-bio, 10: Times of India Back to Main													
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