

Machine Learning and the Formalisation Of Mathematics: Research Challenges

Lawrence C Paulson FRS

1. Introducing ALEXANDRIA

Thank you, ERC!

Mathematicians are fallible

Look at the footnotes on a **single page**
(118) of Jech's *The Axiom of Choice*

¹ The result of Problem 11 contradicts the results announced by Levy [1963b]. Unfortunately, the construction presented there **cannot be completed**.

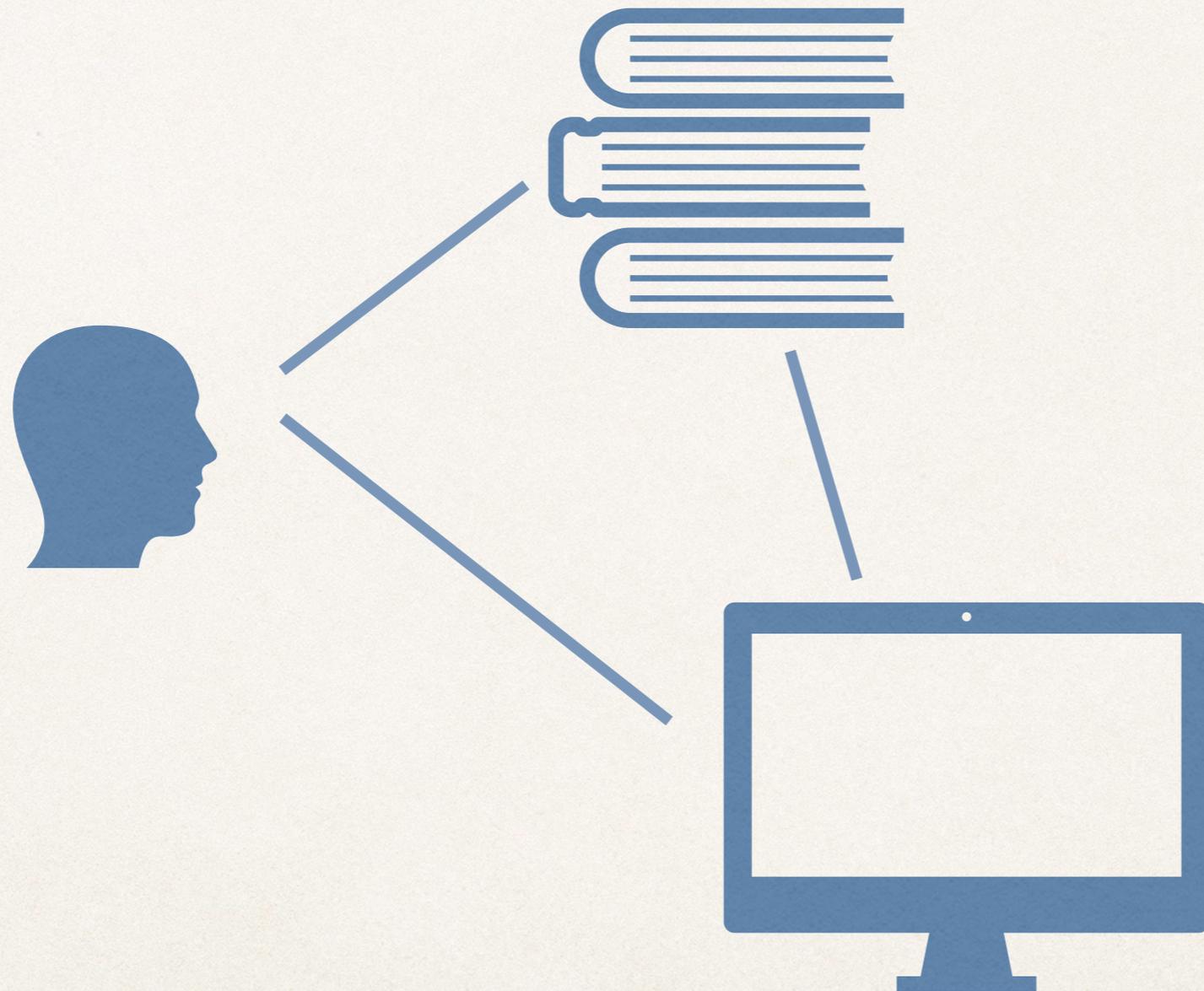
² The transfer to ZF was also claimed by Marek [1966] but the outlined method **appears to be unsatisfactory** and has not been published.

³ A contradicting result was **announced and later withdrawn** by Truss [1970].

⁴ The example in Problem 22 is a counterexample to another condition of Mostowski, who conjectured its sufficiency and singled out this example as a test case.

⁵ The independence result contradicts the claim of Felgner [1969] that the Cofinality Principle implies the Axiom of Choice. **An error has been found** by Morris (see Felgner's corrections to [1969]).

We aim to link people, formal proofs and traditional mathematics



- ❖ Funded by the European Research Council (2017–22)
- ❖ Four postdoctoral researchers:
 - ❖ one Isabelle engineer (*Wenda Li*)
 - ❖ two professional mathematicians (*Angeliki Koutsoukou-Argyraki* and *Anthony Bordg*)
 - ❖ an expert on natural language / machine learning / information retrieval (*Yiannos Stathopoulos*)

What have we been up to?

Building libraries of advanced mathematics:
algebra, analysis,
probability theory...

Writing verified
computer algebra tools

Natural language search
for library theorems

Conjecture synthesis via ML

Aiming to support the
re-use of proof fragments

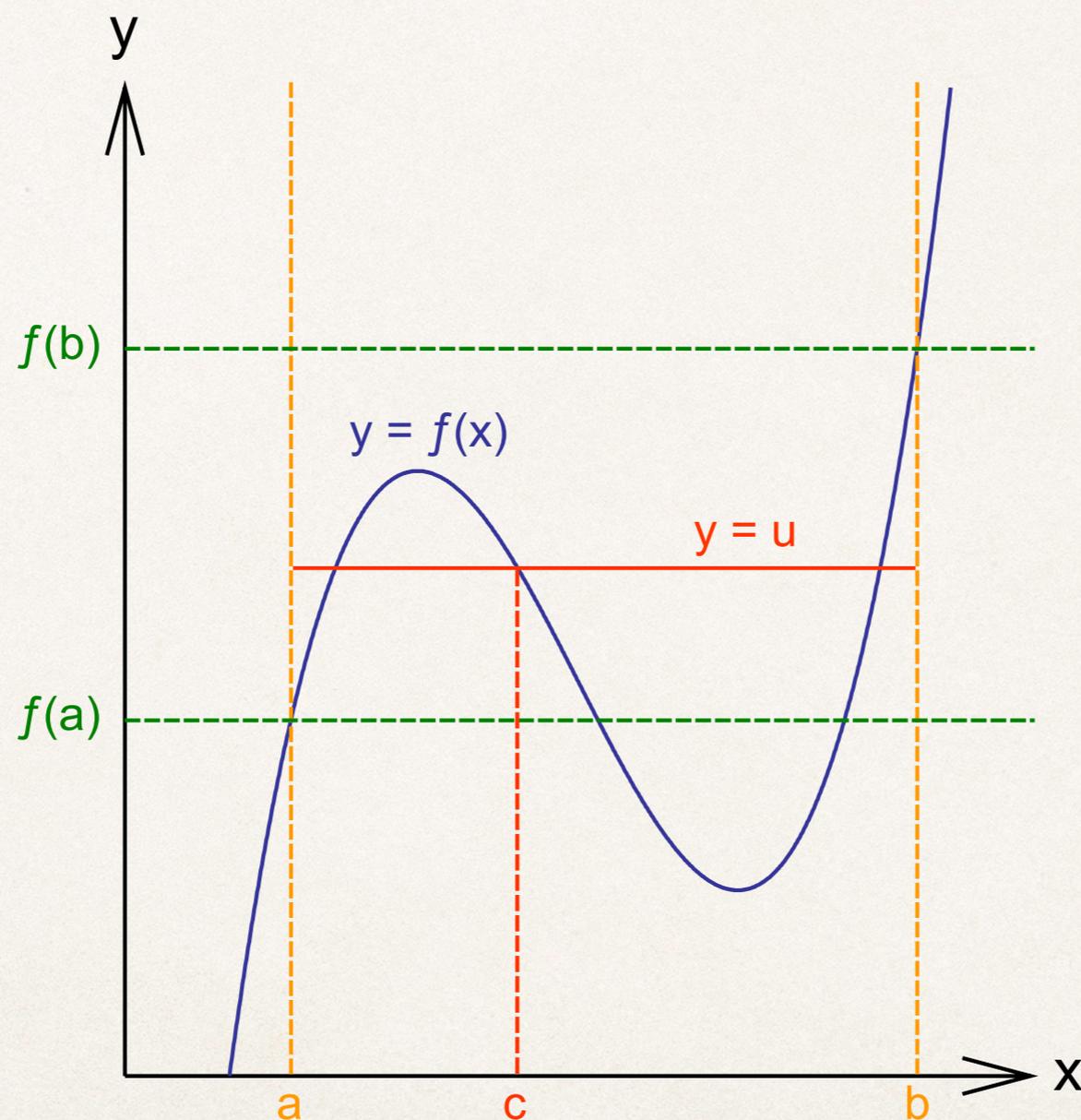
2. Structured Proofs

Tactic proofs: fit only for machines

Intermediate value theorem

```
let IVT = prove(
  `!f a b y. a <= b /\
    (f(a) <= y /\ y <= f(b)) /\
    (!x. a <= x /\ x <= b ==> f cont1 x)
  ==> (?x. a <= x /\ x <= b /\ (f(x) = (y:real)))` THEN
  REPEAT GEN_TAC THEN DISCH_THEN(MP_TAC o SPEC `x:real`) THEN ASM_REWRITE_TAC[] THEN DISCH_TAC THEN
  DISCH_THEN(CONJUNCTS_THEN2 ASSUME_TAC UNDISCH_TAC) THEN
  (CONJUNCTS_THEN2 MP_TAC STRIP_ASSUME_TAC) THEN
  CONV_TAC CONTRAPOS_CONV THEN
  DISCH_THEN(ASSUME_TAC o CONV_RULE NOT_EXISTS_CONV) THEN
  (MP_TAC o C SPEC BOLZANO_LEMMA) THEN
  `(u,v). a <= u /\ u <= v /\ v <= b` THEN
  CONV_TAC(ONCE_DEPTH_CONV GEN_BETA_CONV) THEN
  W(C SUBGOAL_THEN (fun t => REWRITE_TAC[t]) o DISCH_THEN) THEN
  funpow 2 (fst o dest_imp) o snd) THEN
  [ALL_TAC;
  DISCH_THEN(MP_TAC o SPEC [a:real; b:real]) THEN
  ASM_REWRITE_TAC[REAL_LE_refl] THEN
  CONJ_TAC THENL
  [MAP_EVERY X_GEN_TAC [u:real; v:real] THEN
  CONV_TAC CONTRAPOS_CONV THEN
  STRIP_TAC THEN
  MAP_EVERY ASM_CASES_TAC [u:real; v:real] THEN
  DISJ_CASES_TAC(SPECL [y:real; f(v:real)] THEN
  ASM_REWRITE_TAC[] THENL
  [MATCH_MP_TAC REAL_LE_TRANS THEN
  EXISTS_TAC `w:real`; EXISTS_TAC `v:real` THEN
  ALL_TAC] THEN
  X_GEN_TAC `x:real` THEN
  [ALL_TAC;
  EXISTS_TAC `&1` THEN
  MAP_EVERY X_GEN_TAC [u:real; v:real] THEN
  REPEAT STRIP_TAC THEN
  REWRITE_TAC[] THEN
  [EXISTS_TAC `u:real`; EXISTS_TAC `v:real`] THEN
  ASM_REWRITE_TAC[] THEN
  DISCH_THEN(MP_TAC o SPEC `u - x`) THEN
  ASM_REWRITE_TAC[REAL_NOT_LT; REAL_LE_NEG; real_sub; REAL_LE_RADD]]];;
```

Where's the intuition?



By Kpengboy (Own work, based off Intermediatevaluetheorem.png), via Wikimedia Commons

Or again: a HOL Light tactic proof

```
let SIMPLE_PATH_SHIFT_PATH = prove
(`!g a. simple_path g /\ pathfinish g = pathstart g /\
  a IN interval[vec 0,vec 1]
  ==> simple_path(shiftpath a g)` ,
 REPEAT GEN_TAC THEN REWRITE_TAC[simple_path] THEN
 MATCH_MP_TAC(TAUT
  `(a /\ c /\ d ==> e) /\ (b /\ c /\ d ==> f)
  ==> (a /\ b) /\ c /\ d ==> e /\ f`) THEN
 CONJ_TAC THENL [MESON_TAC[PATH_SHIFT_PATH]; ALL_TAC] THEN
 REWRITE_TAC[simple_path; shiftpath; IN_INTERVAL_1; DROP_VEC;
  DROP_ADD; DROP_SUB] THEN
 REPEAT GEN_TAC THEN DISCH_THEN(CONJUNCTS_THEN2 MP_TAC ASSUME_TAC) THEN
 ONCE_REWRITE_TAC[TAUT `a /\ b /\ c ==> d <=> c ==> a /\ b ==> d`] THEN
 STRIP_TAC THEN REPEAT GEN_TAC THEN
 REPEAT(COND_CASES_TAC THEN ASM_REWRITE_TAC[]) THEN
 DISCH_THEN(fun th -> FIRST_X_ASSUM(MP_TAC o C MATCH_MP th)) THEN
 REPEAT(POP_ASSUM MP_TAC) THEN
 REWRITE_TAC[DROP_ADD; DROP_SUB; DROP_VEC; GSYM DROP_EQ] THEN
 REAL_ARITH_TAC);;
```

The same, as a *structured* proof

```
lemma simple_path_shiftpath:
  assumes "simple_path g" "pathfinish g = pathstart g" and a: "0 ≤ a" "a ≤ 1"
  shows "simple_path (shiftpath a g)"
  unfolding simple_path_def
proof (intro conjI impI ballI)
  show "path (shiftpath a g)"
  by (simp add: assms path_shiftpath simple_path_imp_path)
  have *: "∧x y. [[g x = g y; x ∈ {0..1}; y ∈ {0..1}]] ⇒ x = y ∨ x = 0 ∧ y = 1 ∨ x = 1 ∧ y = 0"
  using assms by (simp add: simple_path_def)
  show "x = y ∨ x = 0 ∧ y = 1 ∨ x = 1 ∧ y = 0"
  if "x ∈ {0..1}" "y ∈ {0..1}" "shiftpath a g x = shiftpath a g y" for x y
  using that a unfolding shiftpath_def
  by (force split: if_split_asm dest!: *)
qed
```

Proofs with gaps

a chain of “stepping stones”
from the assumptions to
conclusion

Users can fill these gaps
in *any order*

Structured proofs are necessary!

- ❖ Because formal proofs should *make sense to users*
 - ... reducing the need to **trust** our verification tools
- ❖ For *reuse* and eventual *translation* to other systems
- ❖ For *maintenance* (easily fix proofs that break due to changes to definitions... or **automation**)

*With some other systems,
users **avoid** automation for that reason!*

3. Implications for ML

New possibilities for ML with structured proofs

- ❖ Working **locally** within a large proof
- ❖ Looking for just the **next step** (not the whole proof)
- ❖ Proof by analogy
- ❖ Identifying **idioms**

Lots of data

- ❖ About 230K proof lines in Isabelle's maths libraries:
Analysis, Complex Analysis, Number Theory, Algebra
- ❖ Nearly 2.6M proof lines in the *Archive of Formal Proofs*
(not all mathematics though)
- ❖ Hundreds of different authors: diverse styles and topics

Lots of structured “chunks”

- ❖ Structured proof fragments contain *explicit assertions* and **context elements** that could drive learning
- ❖ These might relate to *natural mathematical steps*
 - ❖ Proving a function to be continuous
 - ❖ Getting a ball around a point within an open set
 - ❖ Covering a compact set with finitely many balls

Where does prior work fit in?

- ❖ *TacticToe*, etc., aim to prove theorems automatically within the tactic paradigm, also predicting (just) the **next tactic**
- ❖ Gauthier et al. work on *statistical conjecturing* attempts term and formula synthesis

There's already a trend towards *incremental proof construction* (as opposed to full proofs)

It is essential to *synthesise terms and formulas*

Even tactics take arguments

Structured proofs mostly consist of explicit formulas

4. *A Few Typical Proof Idioms*

Inequality chains

```
have " $|X_m * Y_m - X_n * Y_n| = |X_m * (Y_m - Y_n) + (X_m - X_n) * Y_n|$ "
  unfolding mult_diff_mult ..
also have " $\dots \leq |X_m * (Y_m - Y_n)| + |(X_m - X_n) * Y_n|$ "
  by (rule abs_triangle_ineq)
also have " $\dots = |X_m| * |Y_m - Y_n| + |X_m - X_n| * |Y_n|$ "
  unfolding abs_mult ..
also have " $\dots < a * t + s * b$ "
  by (simp_all add: add_strict_mono mult_strict_mono' a b i j *)
finally show " $|X_m * Y_m - X_n * Y_n| < r$ "
  by (simp only: r)
```

typically by the *triangle inequality*

with simple algebraic manipulations

there are hundreds of examples

Simple topological steps

```
have "open (interior I)" by auto
from openE[OF this <x ∈ interior I>]
obtain e where e: "0 < e" "ball x e ⊆ interior I" .
```

```
define U where "U = (λw. (w - ξ) * g w) ` T"
have "open U" by (metis oimT U_def)
moreover have "0 ∈ U"
  using <ξ ∈ T> by (auto simp: U_def intro: image_eqI [where x = ξ])
ultimately obtain ε where "ε > 0" and ε: "cball 0 ε ⊆ U"
  using <open U> open_contains_cball by blast
```

a neighbourhood around a point within an open set

many similar *but not identical* instances

Summations

```
have "real (Suc n) *R S (x + y) (Suc n) = (x + y) * (∑i≤n. S x i * S y (n - i))"
  by (metis Suc.hyps times_S)
also have "... = x * (∑i≤n. S x i * S y (n - i)) + y * (∑i≤n. S x i * S y (n - i))"
  by (rule distrib_right)
also have "... = (∑i≤n. x * S x i * S y (n - i)) + (∑i≤n. S x i * y * S y (n - i))"
  by (simp add: sum_distrib_left ac_simps S_comm)
also have "... = (∑i≤n. x * S x i * S y (n - i)) + (∑i≤n. S x i * (y * S y (n - i)))"
  by (simp add: ac_simps)
also have "... = (∑i≤n. real (Suc i) *R (S x (Suc i) * S y (n - i)))
  + (∑i≤n. real (Suc n - i) *R (S x i * S y (Suc n - i)))"
  by (simp add: times_S Suc_diff_le)
also have "(∑i≤n. real (Suc i) *R (S x (Suc i) * S y (n - i)))
  = (∑i≤Suc n. real i *R (S x i * S y (Suc n - i)))"
  by (subst sum.atMost_Suc_shift) simp
also have "(∑i≤n. real (Suc n - i) *R (S x i * S y (Suc n - i)))
  = (∑i≤Suc n. real (Suc n - i) *R (S x i * S y (Suc n - i)))"
  by simp
also have "(∑i≤Suc n. real i *R (S x i * S y (Suc n - i)))
  + (∑i≤Suc n. real (Suc n - i) *R (S x i * S y (Suc n - i)))
  = (∑i≤Suc n. real (Suc n) *R (S x i * S y (Suc n - i)))"
  by (simp flip: sum.distrib scaleR_add_left of_nat_add)
also have "... = real (Suc n) *R (∑i≤Suc n. S x i * S y (Suc n - i))"
  by (simp only: scaleR_right.sum)
finally show "S (x + y) (Suc n) = (∑i≤Suc n. S x i * S y (Suc n - i))"
  by (simp del: sum.cl_ivl_Suc)
```

Painful, yet the steps of that proof are routine!

the distributive law $(x + y)z = xz + yz$

the distributive law $x \sum_{i \leq n} a_n = \sum_{i \leq n} xa_n$

the distributive law $\sum_{i \leq n} (a_n + b_n) = \sum_{i \leq n} a_n + \sum_{i \leq n} b_n$

Shifting the index of summation and deleting a zero term

Change-of-variables is also common in such proofs

Can't at least some of these steps be learned from similar previous proofs?

So, an idea: link common “utility lemmas”
to natural language concepts?

... then let users supply natural language hints?

This shouldn't require too much laborious
lemma tagging: just a few dozen lemmas
would cover many techniques

But for which sort of user?

- ❖ For *mathematicians*, who need help
 - ❖ to use the proof assistant
 - ❖ to navigate its library
 - ❖ to locate missing material in the mathematical literature and eventually to formalise it

❖ Or *verification engineers*

❖ who need mathematics for an application

❖ but lack expert knowledge

❖ and again need help finding relevant library items?

Some work of ours

- ❖ *SERAPIS : A Concept-Oriented Search Engine* (next talk!)
- ❖ *IsarStep: a dataset for conjecture synthesis* (Wenda Li)
 - ❖ to propose intermediate propositions within structured proofs
 - ❖ via neural sequence-to-sequence models.
 - ❖ Close to 20% accuracy when synthesising intermediate propositions.
 - ❖ Can also capture the relationships between concepts, e.g. sets vs. their members.

Conclusions

- ❖ the formalisation of mathematics, especially into structured proofs, requires a different approach to ML
- ❖ *synthesis* of terms and assertions to *continue* (not necessarily complete) a proof
- ❖ *linking* between informal proof ideas and their formal equivalents
- ❖ *brainstorming* backed by the system's full knowledge