

# Machine Learning for Symbolic Integration Algorithm Selection

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# Overview

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1. Interaction Between Machine Learning and Computer Algebra
  - a) Directing solving problems vs. Algorithm Improvement
  - b) Goals for Symbolic Integration
2. Generating Data
  - a) Current Methods
  - b) New Methods
3. Machine Learning
  - a) LSTMs and TreeLSTMs
  - b) Results

# What is Maple?

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- Mathematics software known as a Computer Algebra System.
- Easily manipulate expressions and perform complex calculations.
- Used in Education, Research, and Industry.
- Has been around since 1988 starting at the University of Waterloo.

# Integration In Maple

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- The int function in Maple is essentially a meta-algorithm.
- The function has a choice of 12 different sub-algorithms available to the user.
- Guards are implemented to prevent trying all the sub-algorithms.
- The first answer that is successful is returned to the user.

```
> int(sin(x), x, method=_RETURNVERBOSE)
["lookup" = -cos(x), "default" = -cos(x), "norman" = - $\frac{2}{1 + \tan\left(\frac{x}{2}\right)^2}$ , "meijerg"
=  $\sqrt{\pi} \left( \frac{1}{\sqrt{\pi}} - \frac{\cos(x)}{\sqrt{\pi}} \right)$ , "risch" = -cos(x), "parallelsch" = -cos(x) - 1,
FAILS = ("gosper", "derivativedivides", "trager", "elliptic", "pseudoelliptic", "parts")]
```

# Motivation - Maple and Machine Learning

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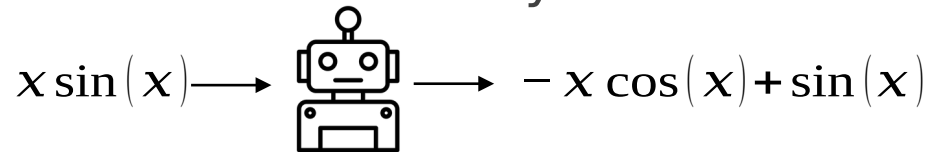
- Computer Algebra refers to the study and development of algorithms and software for manipulating mathematical expressions and other mathematical objects.
- As a Computer Algebra System, Maple should always return the correct answer.
  - Alternatively, Maple shouldn't output anything at all if there is no answer or it cannot compute one!
- Machine Learning has seen many applications in various fields. Computer Algebra is now starting to catch up.
- A problem exists between Computer Algebra and Machine Learning.
  - E.g. I build a model that has 99% accuracy for computing an integral given an expression. Is this acceptable?

# Machine Learning and Integration

- Two approaches:

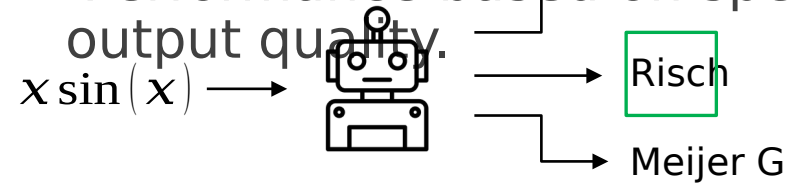
Directly solving a problem

- Compute the result of a task given an input.
- E.g. Given an expression, calculate its integral.
- Performance based on accuracy.



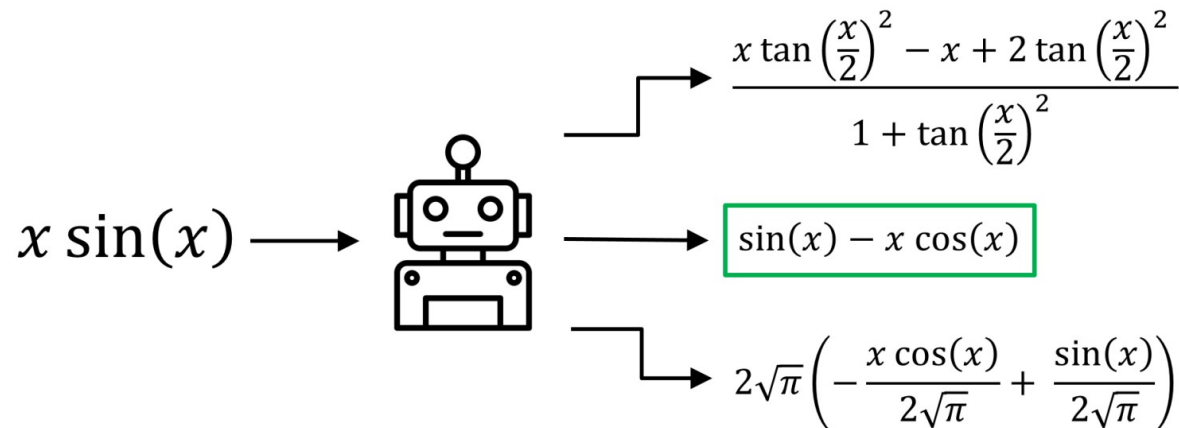
Algorithm Selection

- If an algorithm can make an arbitrary choice, use ML to help guide that choice.
- E.g. Given an expression, which integration rule should we first try?
- Performance based on speed & output quality.



# Objective

- There are two objective functions we can consider when assessing how well a sub-algorithm does.
  - Output length
  - Runtime



- Sub-algorithms selected are the ones that output the smallest length.
  - Could be that a sub-algorithm was successful but gave a longer answer so we consider that a failure.
- Sub-algorithms are not mutually exclusive.

# Comparing to Maple

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- We will train an ML model to select the sub-algorithm that outputs the shortest length.
- The model will be compared to the meta-algorithm that Maple uses.
- Goals of the project:
  - A large quantity and rich variety of data.
  - The model should generalise outside the training data.
  - The output of the model should select sub-algorithms with smaller lengths compared to the meta-algorithm.



# Creating a Labelled Dataset

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*Deep Learning for Symbolic Mathematics* - Lample G, Charton F  
(Meta AI research)

- FWD: Integrate an expression  $f$  through a CAS to get  $F$  and add the pair  $(f, F)$  to the dataset.
- BWD: Differentiate an expression  $f$  to get  $f'$  and add the pair  $(f', f)$  to the dataset.
- IBP: Given two expressions  $f$  and  $g$ , calculate  $f'$  and  $g'$ . If  $\int f'g$  is known then the following holds (integration-by-parts):

$$\int f g' = f g - \int f' g.$$

Thus we add the pair  $(f g', f g - \int f' g)$  to the dataset.

# The Risch Algorithm

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- The algorithm takes as input an expression and outputs its anti-derivative.
  - Anti-derivative must be elementary, otherwise it won't output anything.
- Let  $F$  where  $F$  is a field, and  $E$  is an elementary extension of  $F$ . Let  $f \in E$ . Then the Risch algorithm does the following:

$$\int \frac{a}{b} = \int P + \int \frac{R}{b} \longrightarrow \begin{array}{l} \text{Rational} \\ \text{Part} \end{array}$$

$\downarrow$   
Polynomial Part

# Substitution Rule

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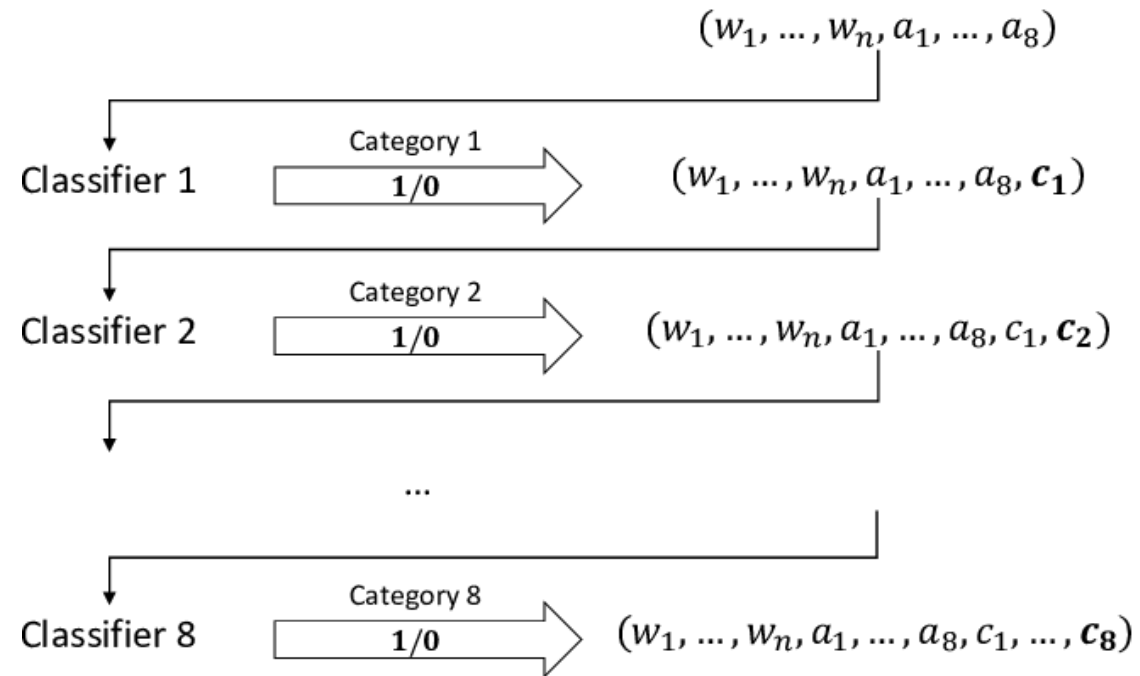
- Integration technique from calculus
- 
- 
- Requires a dataset of integrable expressions

# Algorithm Selection Using Machine Learning

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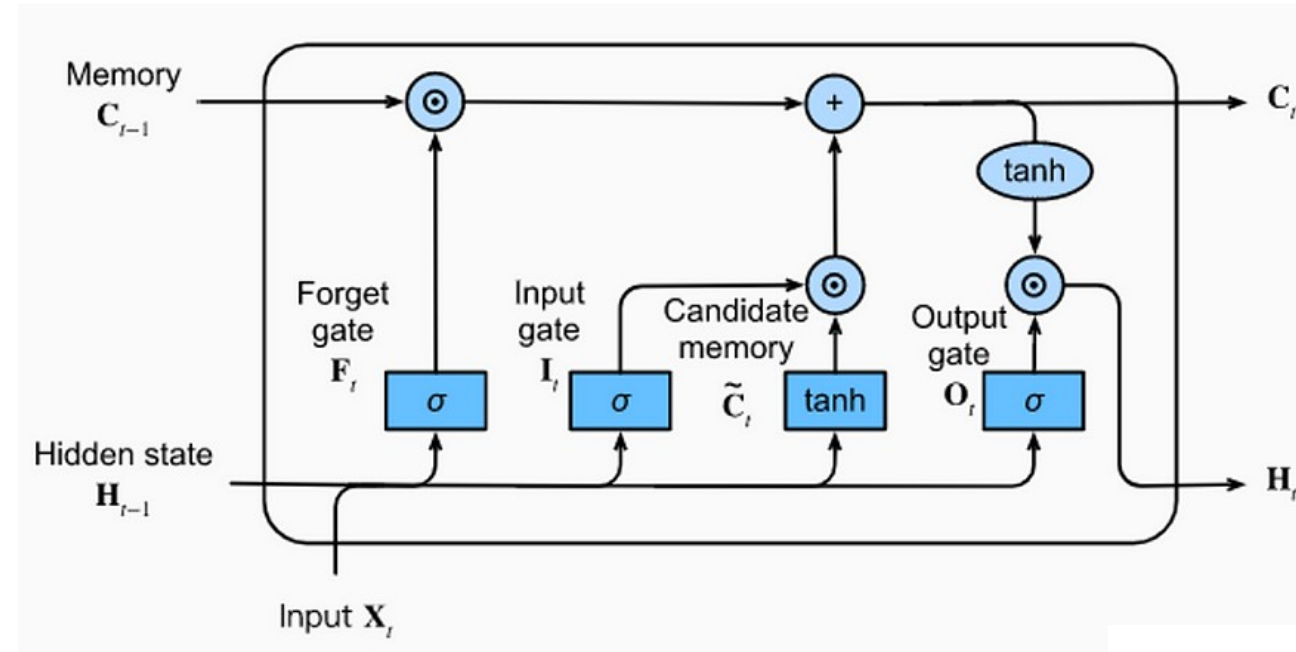
# Multilabel Classification

- Multilabel classification is a type of classification task where each instance can be assigned to multiple classes or labels simultaneously.
- Examples in other fields:
  - Music Genre Classification.
  - Image Segmentation.
  - Sentiment Analysis.
- Multiple integration sub-algorithms can produce the optimal length.
- Two approaches: Binary Relevance and Classifier Chains



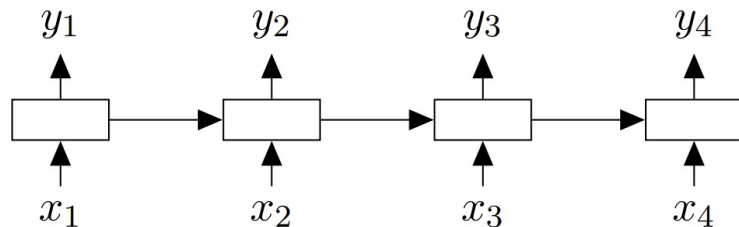
# LSTMs

- LSTM = Long Short-Term Memory
- A Neural Network architecture for handling sequence data (text, time series, etc.)
- Able to remember information far in the past (Long term memory) as well as use the information near the current step (short term memory)
- Performs much better than vanilla neural networks for tasks such as text classification, language translation, and time series predictions



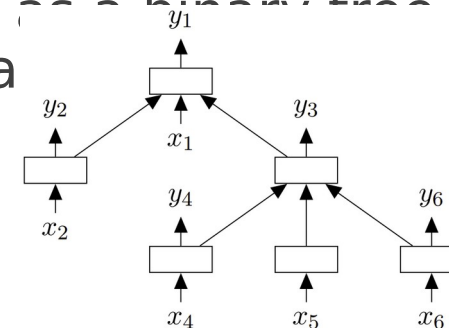
# LSTM

- Used for sequential data processing
- The memory cell updates are dependent on the input at the current timestep and the hidden state of the previous timestep
- Math expressions can be represented as a sequence of tokens to fit the LSTM framework



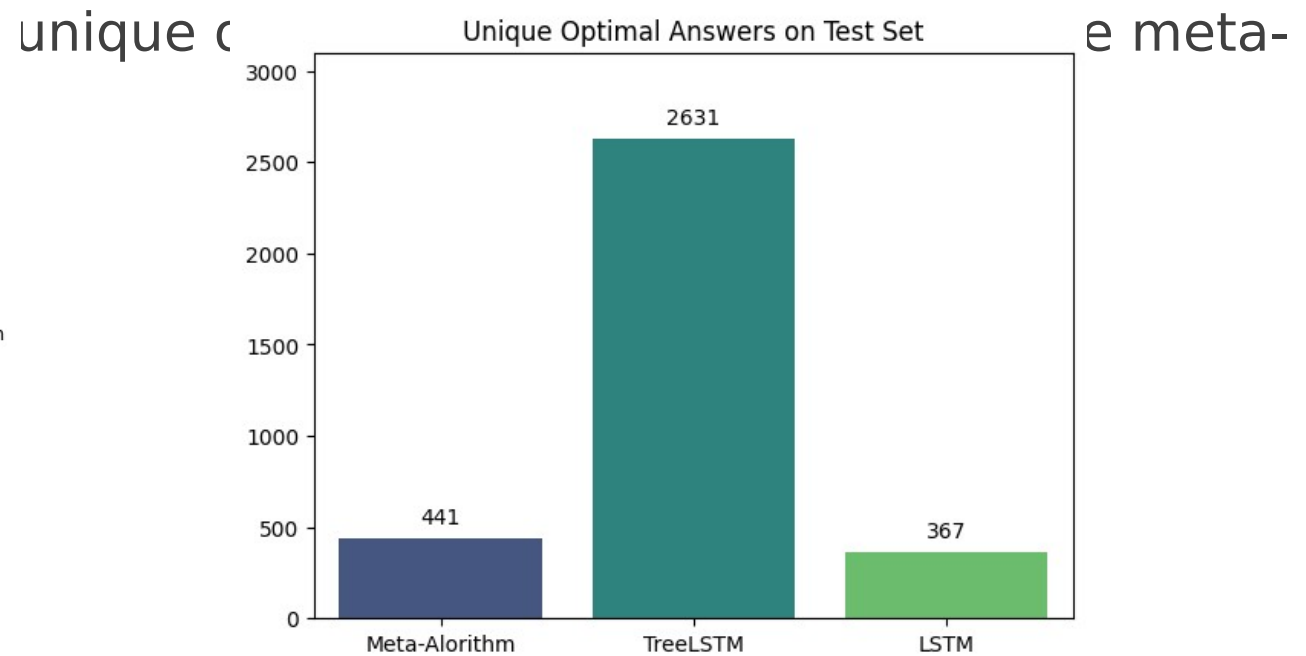
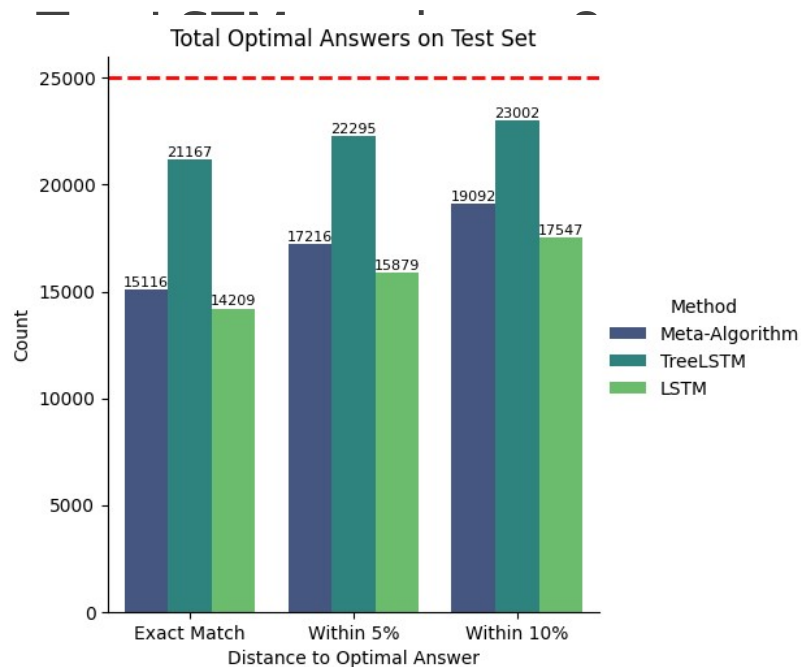
# TreeLSTM

- Used for tree-based data processing
- The memory cell updates are dependent on the states of possibly many child units.
- A regular LSTM can be seen as a TreeLSTM where each cell has one child to form a chain.
- Math expressions can be represented as a binary tree to fit the TreeLSTM framework



# Results on Generated Test Data

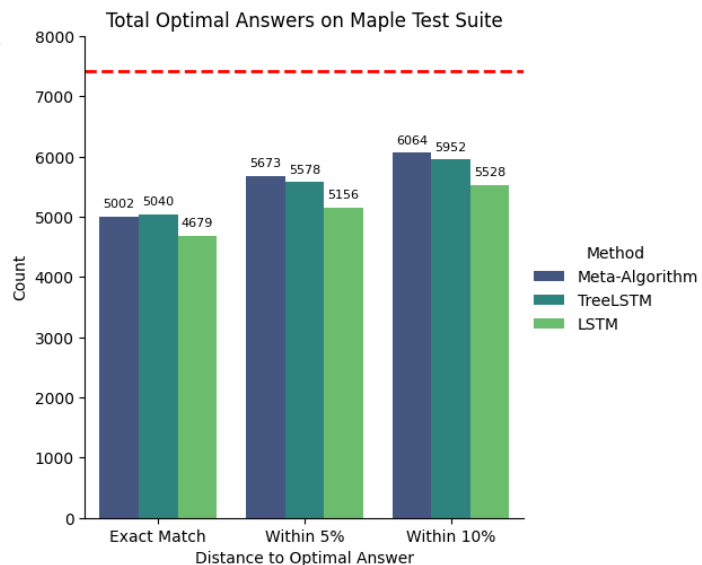
- We generate 15,000 test cases from the FWD, BWD, and IBP methods
- The TreeLSTM and LSTM outperform Maple's meta-algorithm in predicting total optimal answers



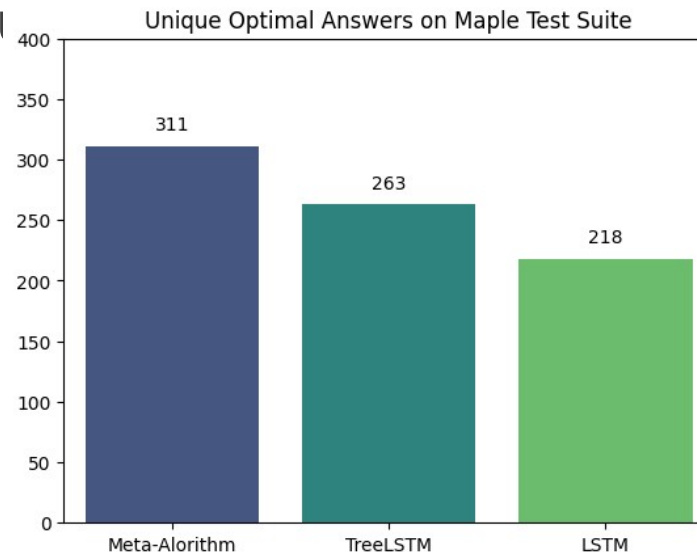


# Results on Maple Test Suite

- Maple has an in-house test suite for their integration function.
  - Consists of 47,500 examples
  - Model is only trained on **elementary** integrable expressions → Only use 7450 examples
- TreeLSTM slightly outperforms meta-algorithm on total optimal answers
- Meta-



Number in unique



```
> f := 1 - x*cos(cos(x))*sin(x) + sin(cos(x))
```

$$f := 1 - x \cos(\cos(x)) \sin(x) + \sin(\cos(x))$$

```
> # Maple's answer
```

```
int(f, x)
```

$$\frac{x + x \tan\left(\frac{x}{2}\right)^2 + x \tan\left(\frac{1 - \tan\left(\frac{x}{2}\right)^2}{2\left(1 + \tan\left(\frac{x}{2}\right)^2\right)}\right)^2 + x \tan\left(\frac{x}{2}\right)^2 \tan\left(\frac{1 - \tan\left(\frac{x}{2}\right)^2}{2\left(1 + \tan\left(\frac{x}{2}\right)^2\right)}\right)^2 + 2x \tan\left(\frac{1 - \tan\left(\frac{x}{2}\right)^2}{2\left(1 + \tan\left(\frac{x}{2}\right)^2\right)}\right) + 2x \tan\left(\frac{x}{2}\right)^2 \tan\left(\frac{1 - \tan\left(\frac{x}{2}\right)^2}{2\left(1 + \tan\left(\frac{x}{2}\right)^2\right)}\right)}{\left(1 + \tan\left(\frac{1 - \tan\left(\frac{x}{2}\right)^2}{2\left(1 + \tan\left(\frac{x}{2}\right)^2\right)}\right)^2\right) \left(1 + \tan\left(\frac{x}{2}\right)^2\right)}$$

```
> # ML-suggested answer
```

```
int(f, x, method=risch)
```

$$x + \sin(\cos(x)) x$$

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# Thank you! Questions?