PDF2LATEX

$$\det\left(\frac{\partial^2 F}{\partial x_i \partial x_j}\right) = \sum_{\sigma \in S_n} sgn(\sigma)$$

\det\left(\frac{\partial^2 F}{\partial x_i \partial x_j}
\right) = \sum_{\sigma \in S_n} \text{sgn}(\sigma)

Macéo Ottavy, Mathieu Longatte, Louison Mocq, Ankit Gayen
Supervised by Simon Delamare

43

and the rule is proved that

$$\frac{du^n}{dx} = nu^{n-1}\frac{du}{dx},$$

where n is a positive fraction whose numerator and denominator are integers. This rule has already been used in the solution of numerous exercises.

34. The Derivative of a Constant. Let y = c, where c is a constant. Corresponding to any Δx , $\Delta y = 0$, and consequently

$$\frac{\Delta y}{\Delta x} = 0$$

and

$$\lim_{\Delta x \doteq 0} \frac{\Delta y}{\Delta x} = 0,$$

GI

$$\frac{dy}{dx} = 0$$

The derivative of a constant is zero.

Interpret this result geometrically.

35. The Derivative of the Sum of Two Functions. Let

$$y = u + v$$

where u and vare functions of x. Let Δu , Δv , and Δy be the increments of u, v, and y, respectively, corresponding to the increment Δx .

$$y + \Delta y = u + \Delta u + v + \Delta v$$

$$\Delta y = \Delta u + \Delta v$$

$$\frac{\Delta y}{\Delta x} = \frac{\Delta u}{\Delta x} + \frac{\Delta v}{\Delta x}$$

$$\frac{dy}{dx} = \frac{du}{dx} + \frac{dv}{dx}$$

0

$$\frac{d(u+v)}{dx} = \frac{du}{dx} + \frac{dv}{dx}$$

The derivative of the sum of two functions is equal to the sum of their derivatives. and the rule is proved that

$$\frac{du^*}{dx} = nu^{*-1}\frac{du}{dx},$$

where n is a positive fraction whose numerator and denominator are integers. This rule has already been used in the solution of numerous exercises.

34 The Derivative of a Constant

Let y = c, where c is a constant. Corresponding to any Dx, Dy = 0, and consequently

$$\frac{\Delta y}{\Delta x} = 0,$$

and

$$\lim_{\Delta x \to 0} \frac{\Delta y}{\Delta x} = 0,$$

or

$$\frac{dy}{dx} = 0$$

The derivative of a constant is zero. Interpret this result geometrically.

35 The Derivative of the Sum of Two Functions

Let

$$y = u + v$$

where u and v are functions of x. Let Du, Du, and Dy be the increments of u, v, and y, respectively, corresponding to the increment Dx.

$$y + \Delta y = u + \Delta u + v + \Delta v$$

$$\Delta y = \Delta u + \Delta v$$

$$rac{\Delta y}{\Delta x} = rac{\Delta u}{\Delta x} + rac{\Delta v}{\Delta x}$$

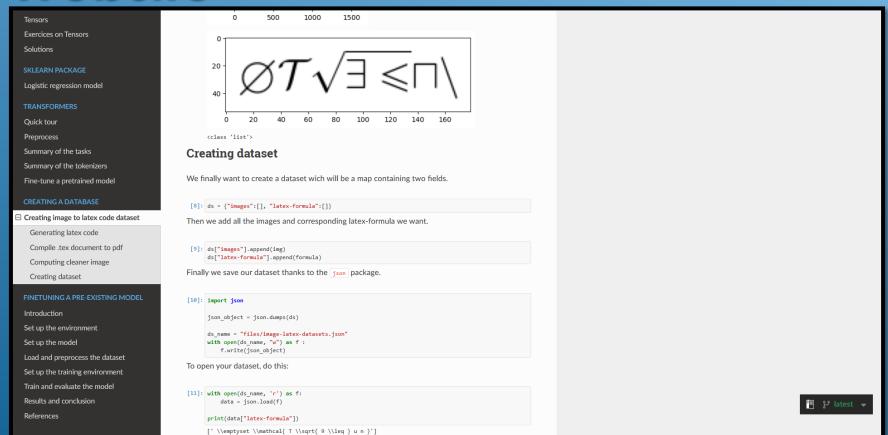
$$\frac{dy}{dx} = \frac{du}{dx} + \frac{dv}{dx},$$

or

$$\frac{d(u+v)}{dx} = \frac{du}{dx} + \frac{dv}{dx}.$$

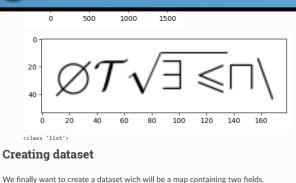
The derivative of the sum of two functions is equal to the sum of their derivatives.

Website



Website

Tensors **Exercices on Tensors** Solutions Logistic regression model Quick tour Preprocess Summary of the tasks Summary of the tokenizers Fine-tune a pretrained model ☐ Creating image to latex code dataset Generating latex code Compile .tex document to pdf Computing cleaner image Creating dataset Introduction Set up the environment Set up the model Load and preprocess the dataset Set up the training environment Train and evaluate the model Results and conclusion References



```
[8]: ds = {"images":[], "latex-formula":[]}
```

Then we add all the images and corresponding latex-formula we want.

```
[9]: ds["images"].append(img)
     ds["latex-formula"].append(formula)
```

Finally we save our dataset thanks to the json package.

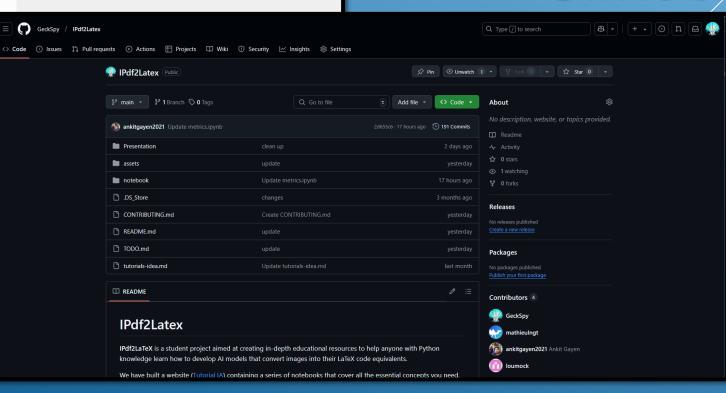
```
[10]: import json
     json_object = json.dumps(ds)
     ds_name = "files/image-latex-datasets.json"
     with open(ds name, "w") as f :
         f.write(json object)
```

To open your dataset, do this:

```
[11]: with open(ds name, 'r') as f:
         data = json.load(f)
```

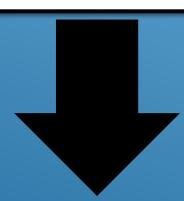
print(data["latex-formula"])

[' \\emptyset \\mathcal{ T \\sqrt{ 9 \\leq } u n }']



Objective: Image to Latex

$$\det\left(\frac{\partial^2 F}{\partial x_i \partial x_j}\right) = \sum_{\sigma \in S_n} sgn(\sigma) \prod_{i=1}^n \frac{\partial^2 F}{\partial x_i \partial x_{\sigma(i)}}$$



Steps

Step 1: Collect data

Easy: Generate dataset from scratch or arxiv or take one from Huggingface

Related tutorials: <u>Basics</u>, <u>Images</u>, <u>Loading</u>, <u>Building</u>

Steps

Step 1: Collect data

Easy: Generate dataset from scratch or arxiv or take one from Huggingface

Related tutorials: <u>Basics</u> <u>Images</u> <u>Loading</u>, <u>Building</u>

Step 2: Chose a relevant architecture

Hard: No obvious structure for Image to text

Hard: Language models require a gigantic amount of data

Related tutorials: <u>Curse of Dimensionality</u>, <u>Summary of the tokenizers</u>, <u>Fine-tune a pretrained model</u>

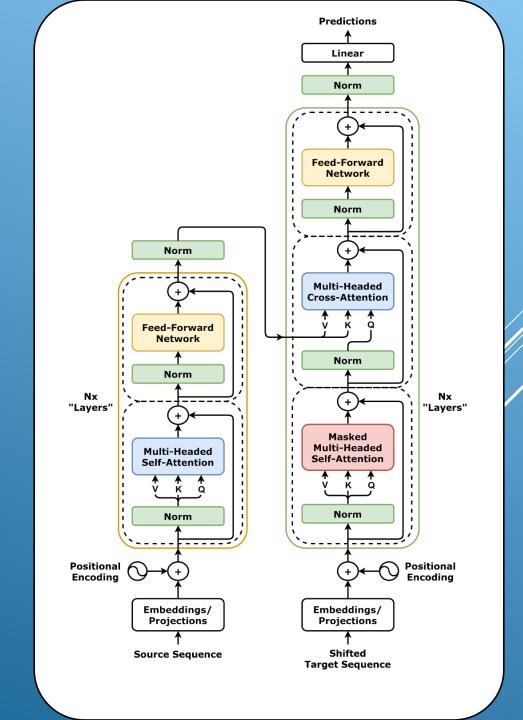
Objectives:

Capture long distance dependency

Enable parallelization in the treatment

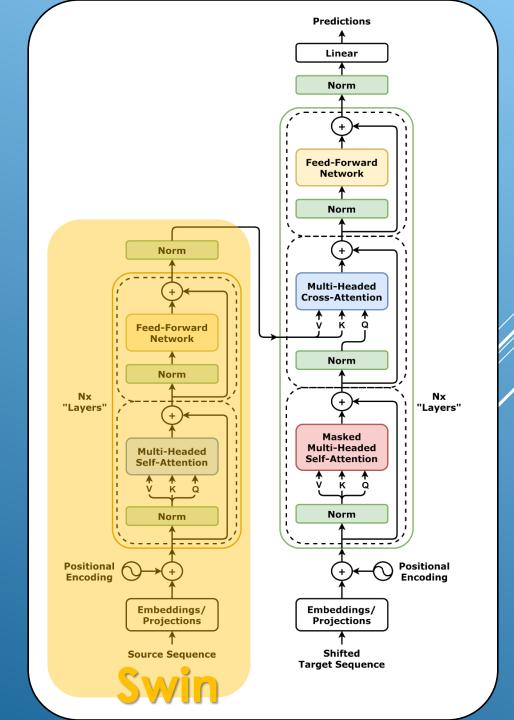
```
Time:
Person:
Ryan
Action:
Object:
a red car
Goal:
Place:

Yesterday,
Ryan
bought
to get to his job
in Paris
```



Encoder:

An encoder in Al transforms input data (e.g., text, images) into a compact numerical representation that captures essential features.



Related tutorials: Transformers

Encoder:

An encoder in AI transforms input data (e.g., text, images) into a compact numerical representation that captures essential features.

Decoder:

A decoder in AI takes this encoded representation and reconstructs it back into a human-interpretable format, such as text or images.

Predictions Linear Norm Feed-Forward Network Multi-Headed Cross-Attention Feed-Forward Network Norm Multi-Headed Multi-Headed Self-Attention Norm Encoding Embeddings/ Embeddings/ **Shifted**

Related tutorials: Transformers

Steps

Step 1: Collect data

Step 2: Chose a relevant architecture

Step 3: Chose metrics to evaluate the model



Related tutorials: Introduction to few NLP metrics

BLEU = Bilingual Evaluation Understudy

We are going to predict mathematical formula

BLEU = Bilingual Evaluation Understudy

We are going to predict mathematical formula

Is it good?

Goal of a metric:

Evaluate efficiently the performances of the model

N-gram = sequence of n consecutive words

We are going to predict mathematical formula

N-gram = sequence of n consecutive words

We are going to predict mathematical formula

N-gram = sequence of n consecutive words

We are going to predict mathematical formula 2-grams:

n-gram precision:

$$\frac{\sum\limits_{\mathcal{C} \in unique\ n-grams} count_clip(\mathcal{C})}{number\ of\ n-grams}$$

n-gram precision:

$$\frac{\sum_{\mathcal{C} \in unique \ n-grams} count_clip(\mathcal{C})}{number \ of \ n-grams}$$

Low *n*: Checks the vocabulary of tokens/words used in the candidate

High n: Checks and ensures proper token order and longer syntactic rules

$$BLEU = BP \cdot \exp\left(\sum_{n=1}^{N} w_n \log p_n\right)$$

n-gram precision decrease exponentially

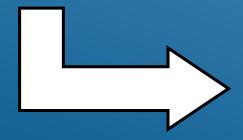
Steps

Step 1: Collect data

Step 2: Chose a relevant architecture

Step 3: Chose metrics to evaluate the model

Step 4: Chose a relevant programming environment



High view library by **Huggingface**: Transformers, Trainer HPC platform: grid5000

Training:

- 30 hours on 4 GPUs on grid5000
- We finetuned 600K parameters among 250M parameters
- 250K examples on the training dataset, during 3 epochs

Results:

Model	Google BLEU	Data Size	Parameters
Im2Latex	0.67	441K	243M
TexTeller	0.77	7.5M	300M
Pix2Text	0.07	100K	25M
Sumen	0.47	6.9M	350M
Our	0.23	250K	250M

Before training

$$\Delta_{+}(x,y) = \sum_{n=-\infty}^{\infty} \tilde{\Delta}(x_0 + nT, \vec{x}; y_0, \vec{y})$$



I don't know if it's a coincidence or not," he said. "I think it's just a matter of time before we get to know each other."

He added: "It's not like we're going to be friends. It's just that we're not going to get along very well.

"We're not friends. We're not even friends. I don't think we'll ever get along. I think we're just going to have to get used to each other. We'll just have to figure out how to get through this together. This article is from the archive of our partner .

After training

$$\Delta_{+}(x,y) = \sum_{n=-\infty}^{\infty} \tilde{\Delta}(x_0 + nT, \vec{x}; y_0, \vec{y})$$



```
{ \Delta } _ { + } ( x , y ) = \sum _ { n = - \infty } ^ { { } } } { \tilde { \Delta \left( } x _ { 0 } + n T , { \vec { x } } ; y _ { 3 } ; 
 { \bf { y } } \right) } . \ , | {% 1 2 3 5 6 7 8 9 0 4 ^ ( ) x p = y d e g f a b c n t v q z ~ r s o i u w " j l k > \( / V _ I + J - : \) \q A
```

After training

$$\Delta_{+}(x,y) = \sum_{n=-\infty}^{\infty} \tilde{\Delta}(x_0 + nT, \vec{x}; y_0, \vec{y})$$



```
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 { \bf { y } } \right) } . \ , | {% 1 2 3 5 6 7 8 9 0 4 ^ ( ) x p = y d e g f a b c n t v q z ~ r s o i u w " j l k > \( / V _ I + J - : \) \q A
```

- Model generalizing well Limitations: Training times, computation power

- Notebooks covering all project

Model generalizing well
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Datasets:

Basics Images Loading Building

Model generalizing well
 Limitations: Training times, computation power

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Datasets: ML:

<u>Basics</u>

<u>Images</u>

<u>Loading</u>

Building

<u>Uses</u>

Training time

Computer's power

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Theory:

<u>GAs</u>

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<u>Datas</u>

Torch

Transformers

Sklearn

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<u>Metrics</u>

<u>Datas</u>

Torch

Transformers

Sklearn

Finetunning;

<u>Set up</u>

Preprocess

Training

Results

Futur/ next steps

- Improve model performance
- Extend to the PDF to Latex tool
- Provide the associated notebooks

IPDF2LATEX

Website: https://tutorial-ia-pe.readthedocs.io/en/latest/

GitHub: https://github.com/GeckSpy/IPdf2Latex

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Supervised by **Simon Delamare**

References

Image-to-LaTeX Converter for Mathematical Formulas and Text

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²Department of Data Science and Artificial Intelligence, Saarland University

{dagu00001, almo00008}@stud.uni-saarland.de

After training

$$N = rac{1}{4eta \left(rac{2}{\sqrt{d_{\perp}}}
ight)^{d_{\perp}}} \int_{y_N}^{y_{end}} rac{(1-y)^{d_{\perp}-1}dy}{y^3 \left(1+(rac{d_{\perp}}{4}-rac{3}{2})y
ight)}$$



```
N = \frac { 1 } { \beta \Lambda ^ { d - 1 } } \int _ { \mu _ { N } } ^ { \nu _ { 0 } }
\alpha _ { s } } { ( \frac 2 { \sqrt { d _ { D } } } ) ^ { 2 } \over { y ^ { 3 } \left(
1 + ( 1 + \frac 3 { 4 } - \frac 1 4 ) y \right) } } \ .%: = , ( ) 1 2 - 3 4 5 7 8 9 6 |
> { ~ \( \sigma \) 0 ^ e p a b c d x r v l s i y h t o k q z n I g f }
```