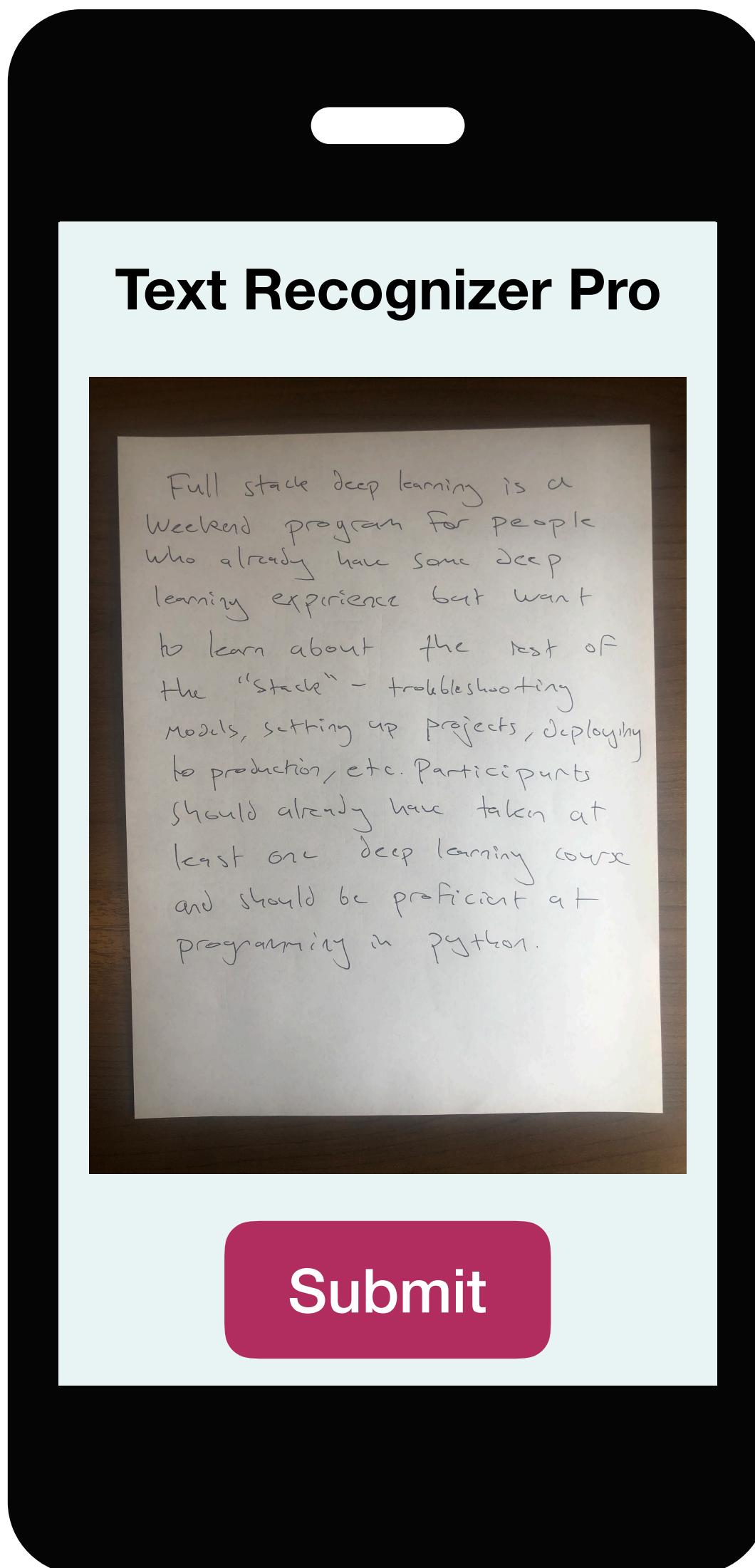


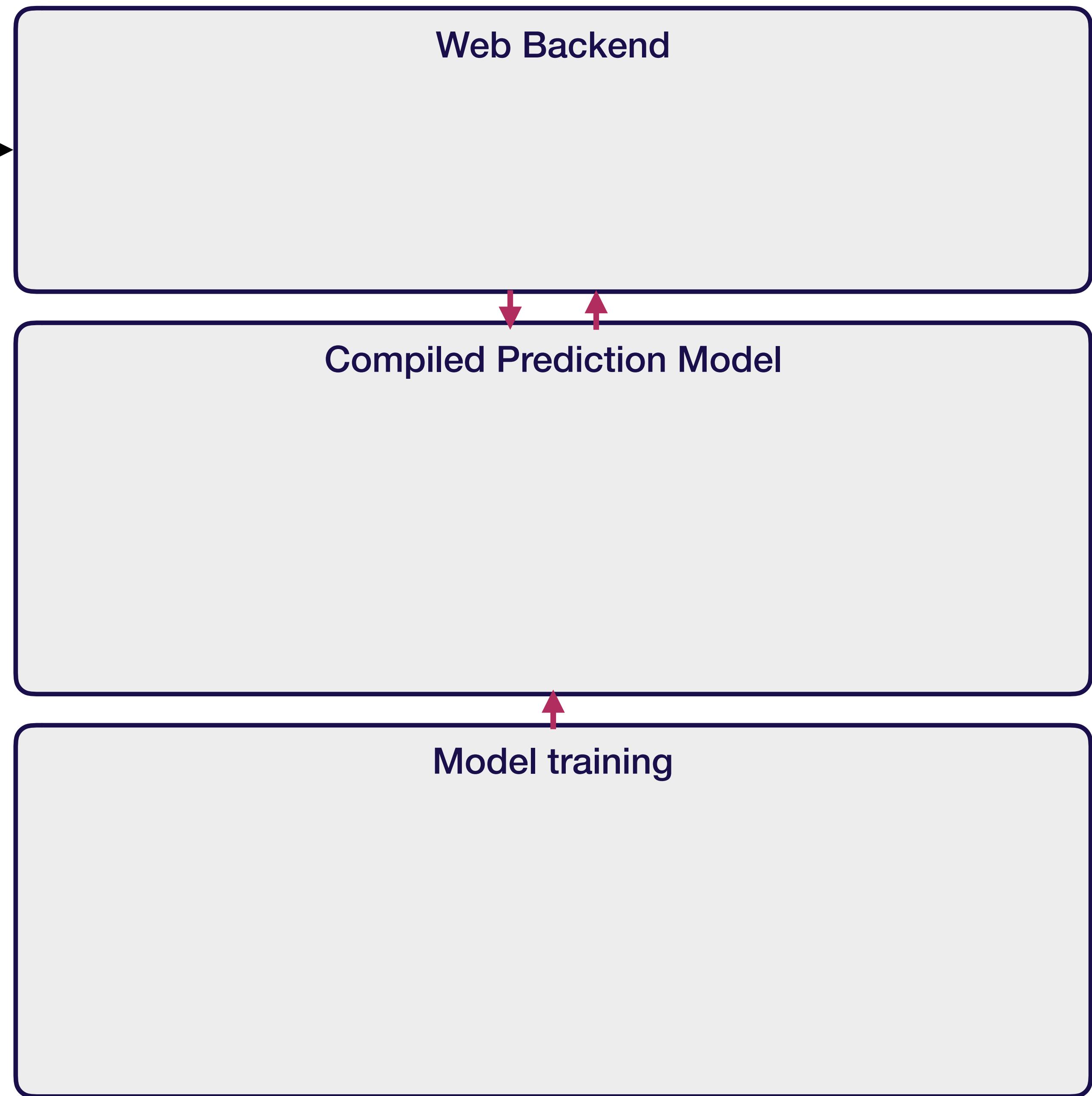
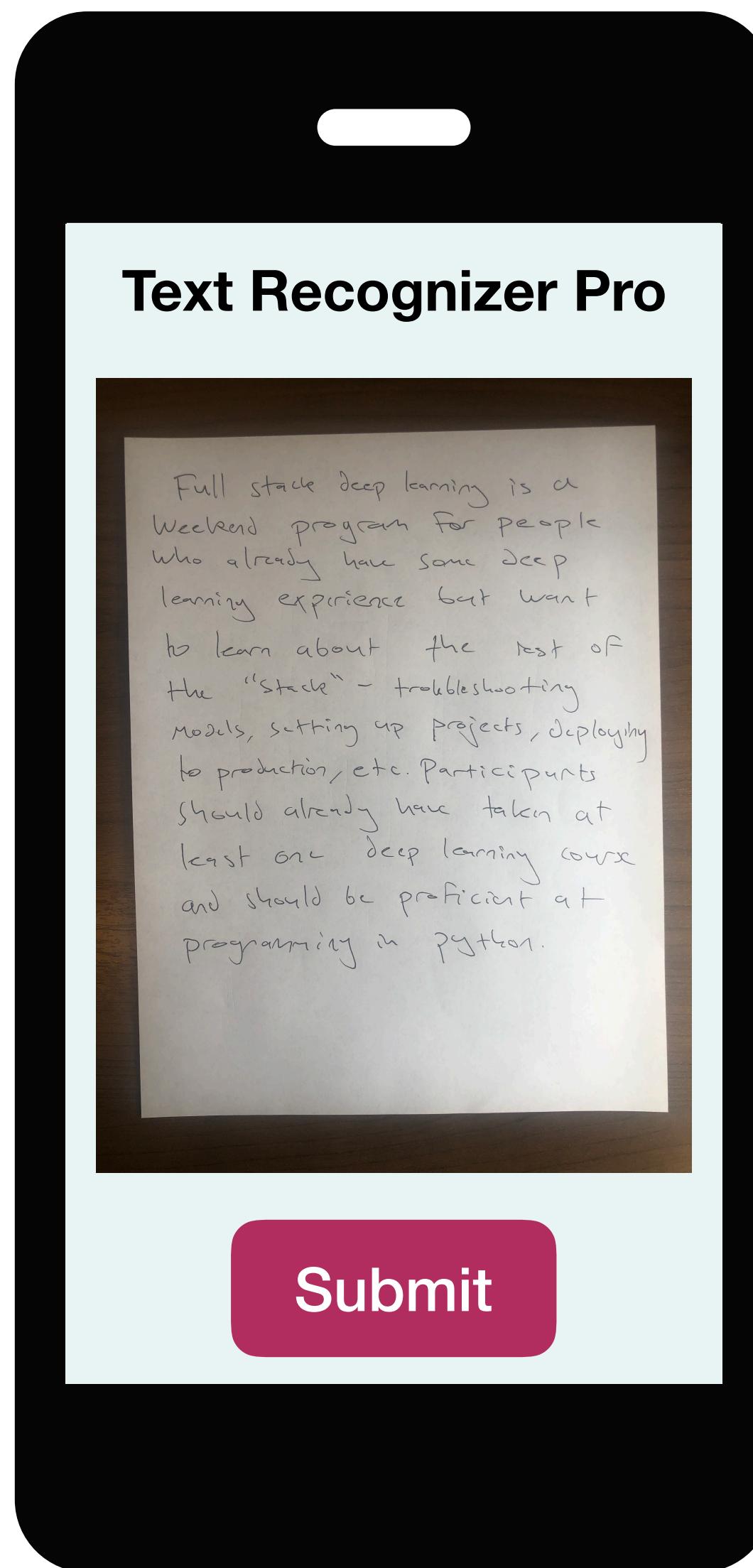
Full Stack Deep Learning

Introduction to the Text Recognizer Lab

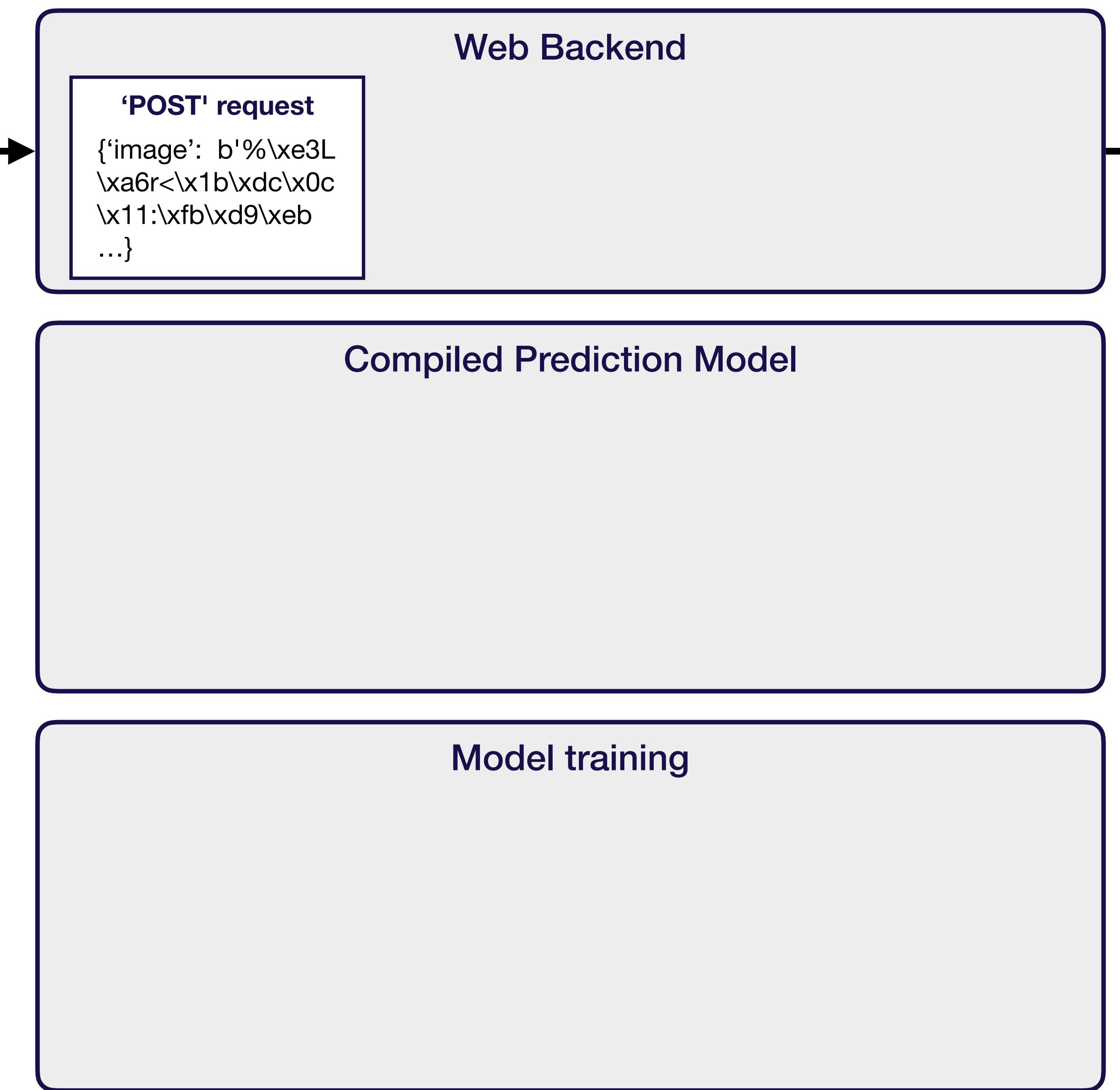
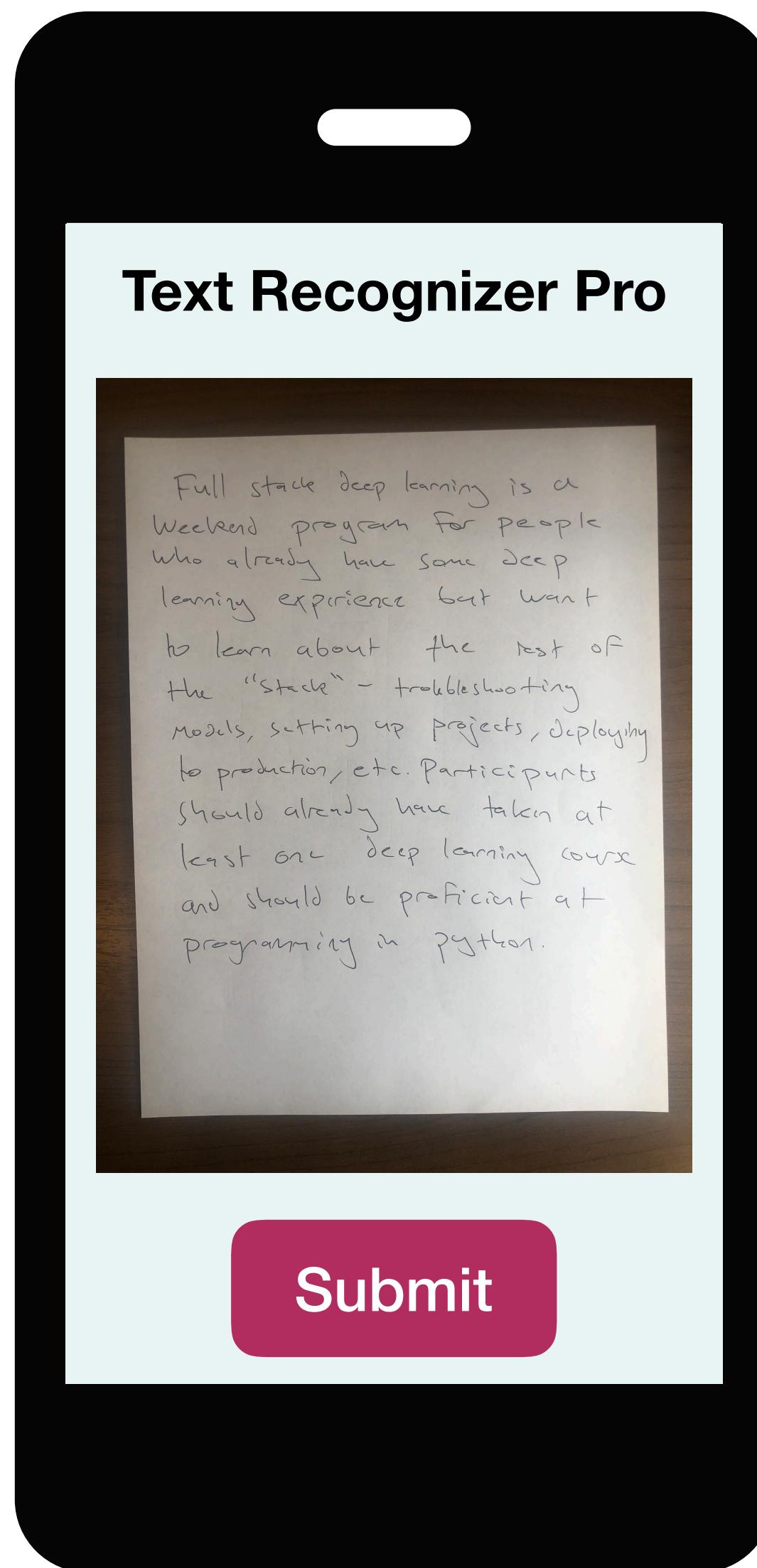
Josh Tobin, Sergey Karayev, Pieter Abbeel



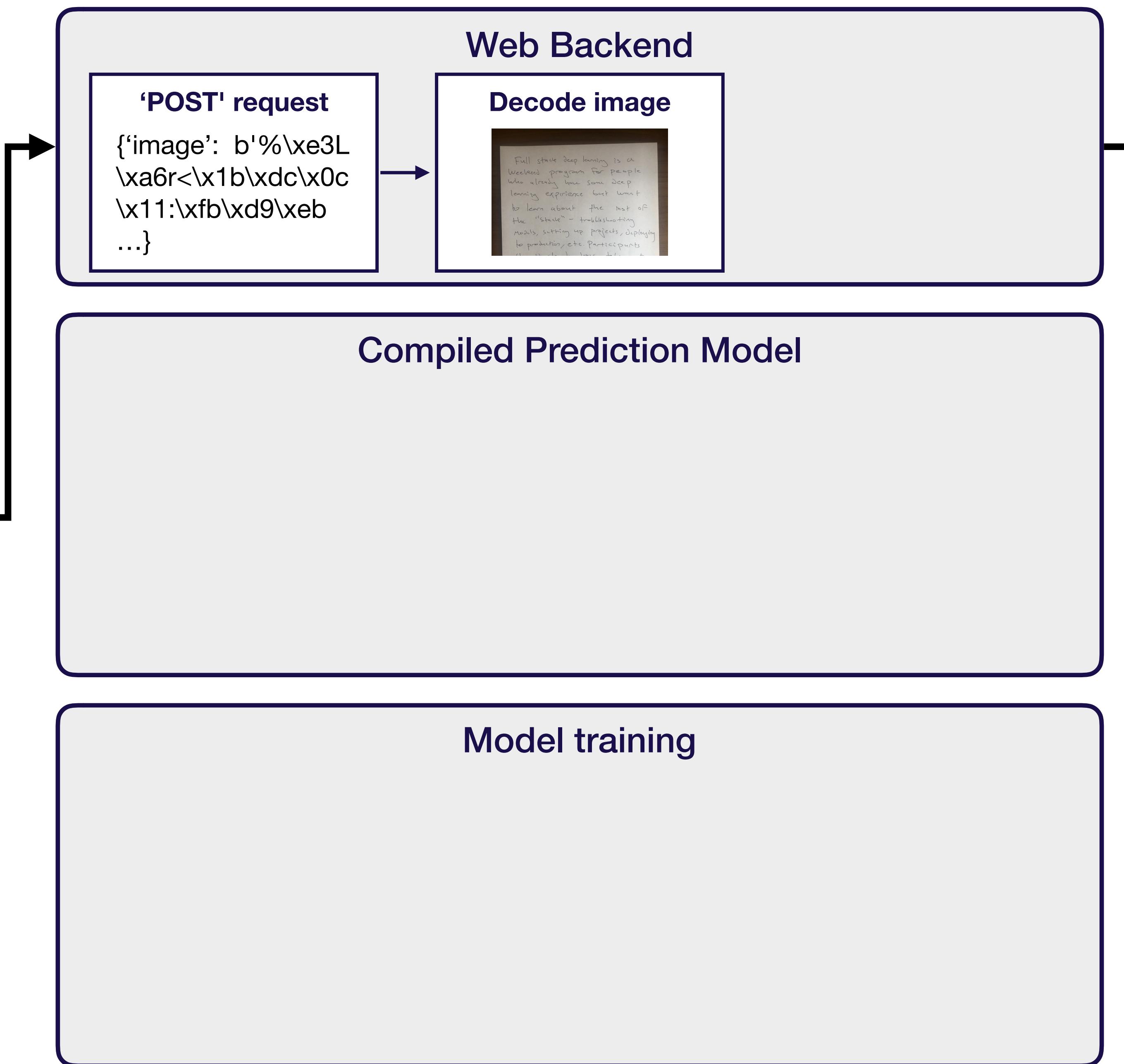
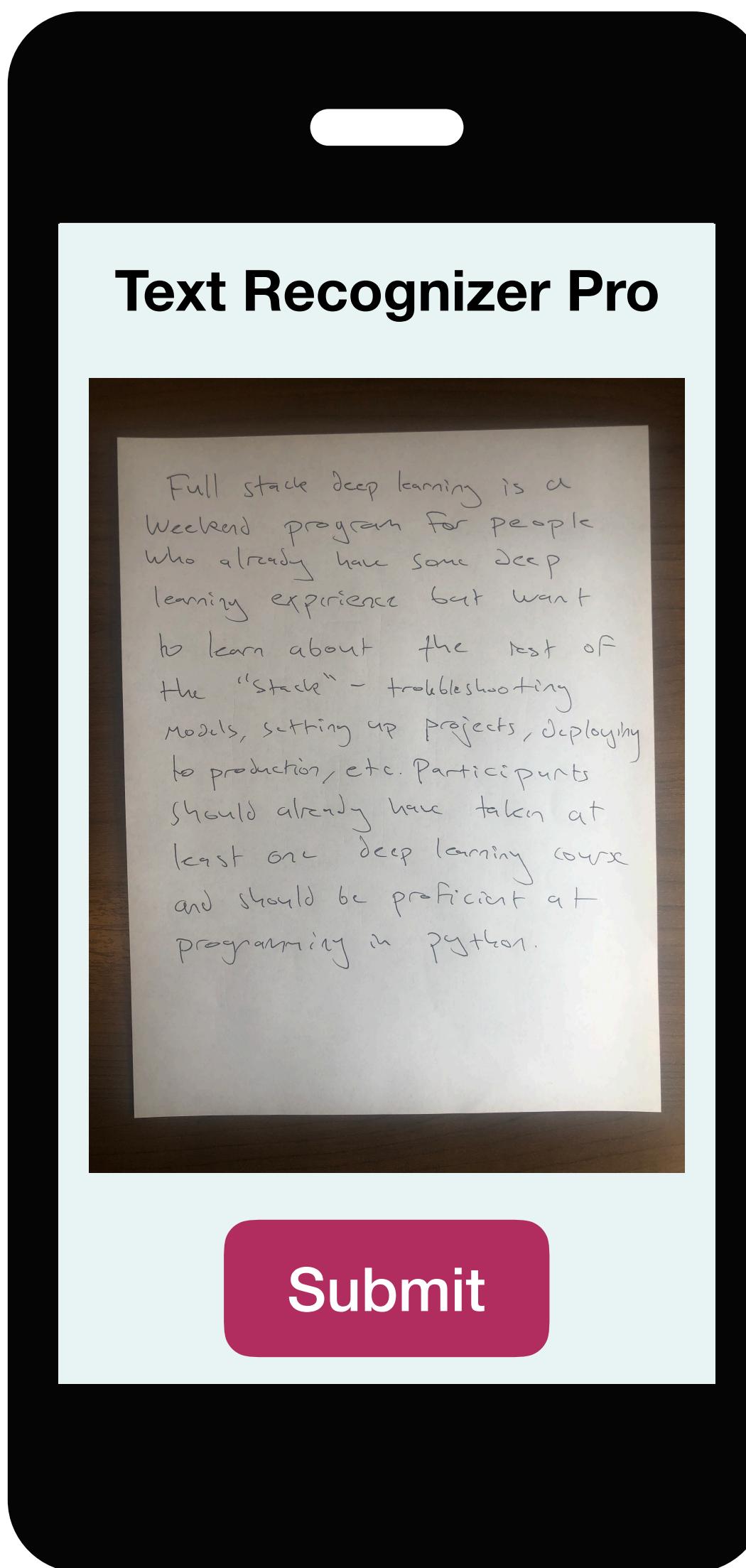
Full stack deep learning is a weekend program for people who already have some deep learning experience but want to learn about the rest of the “stack” ...



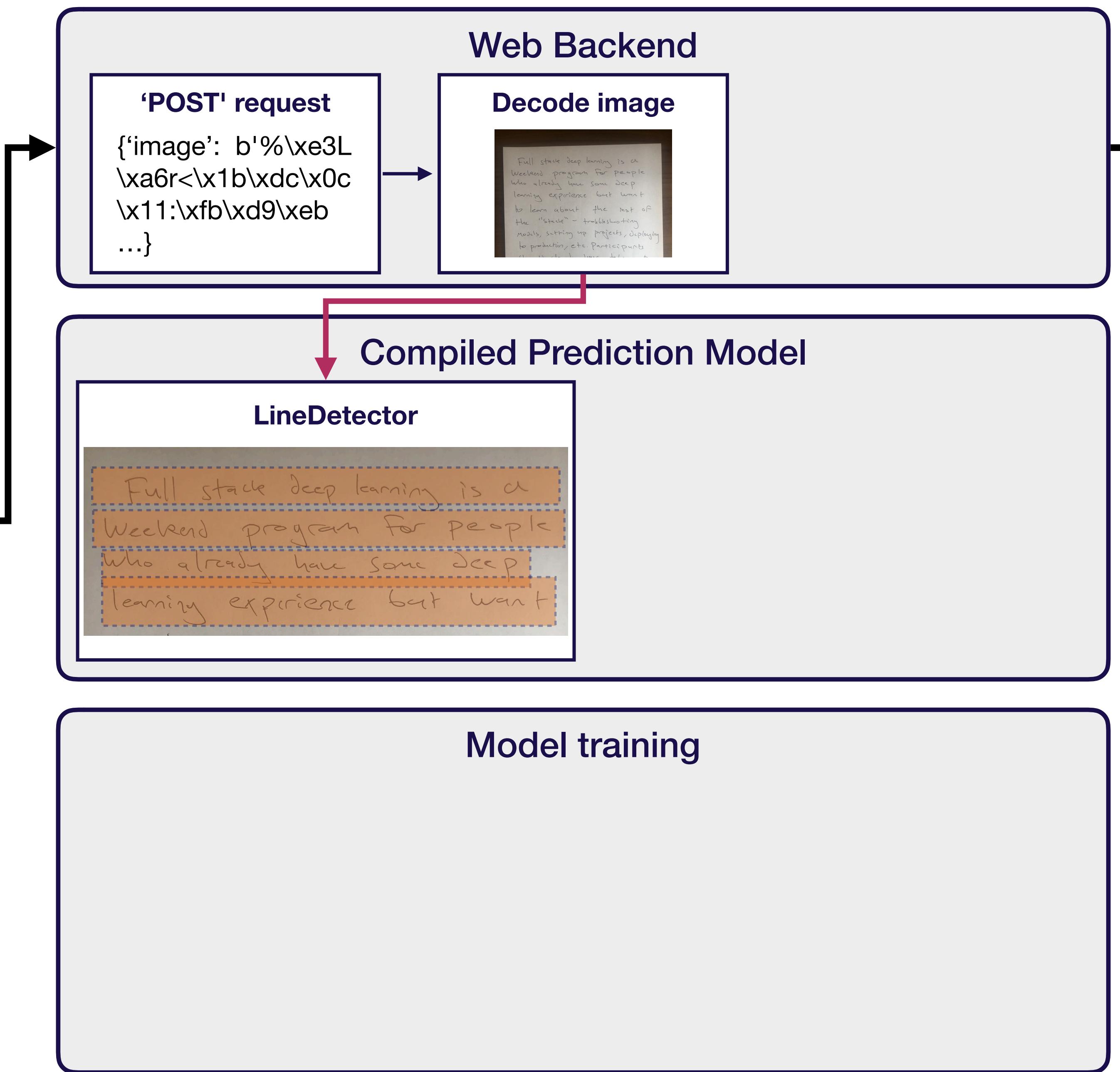
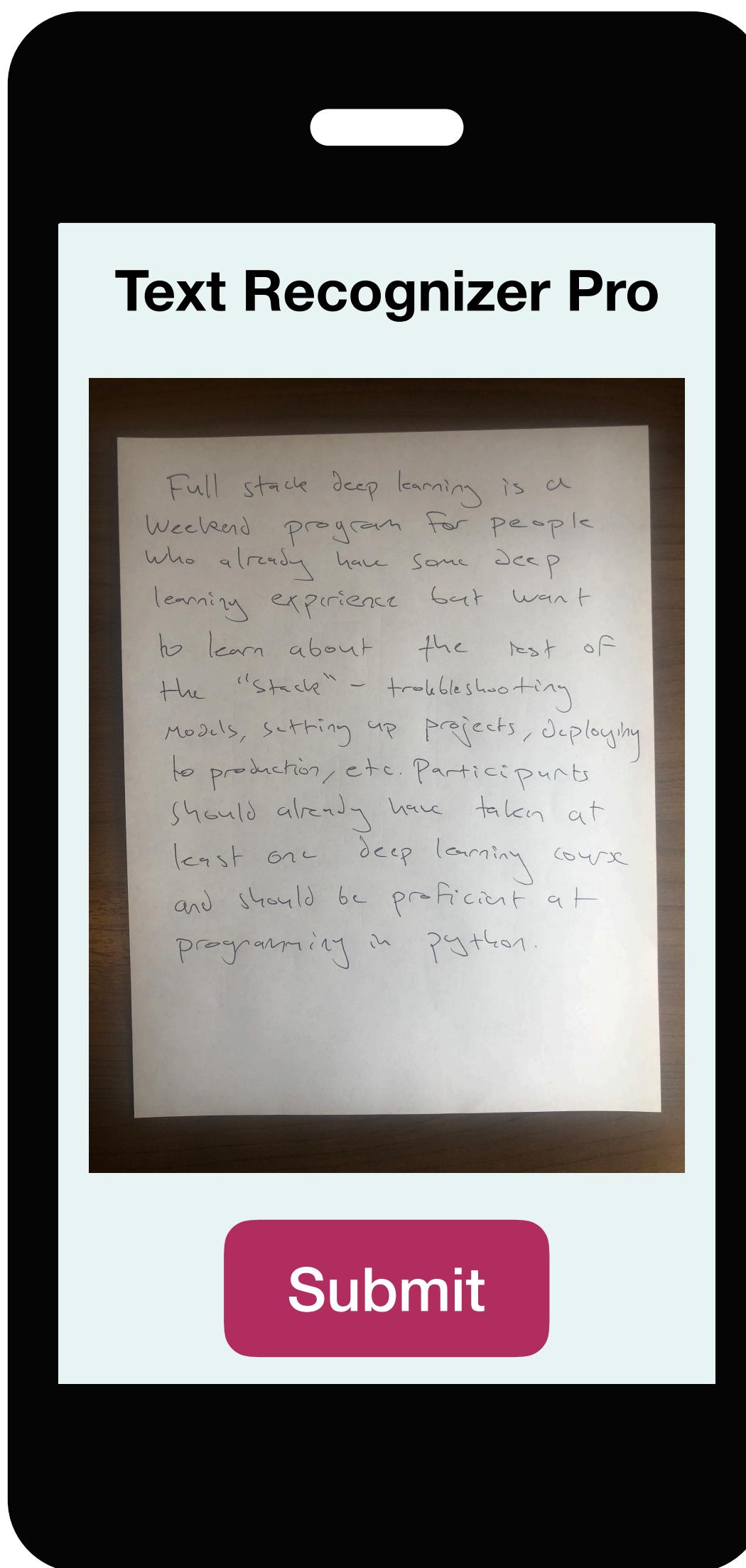
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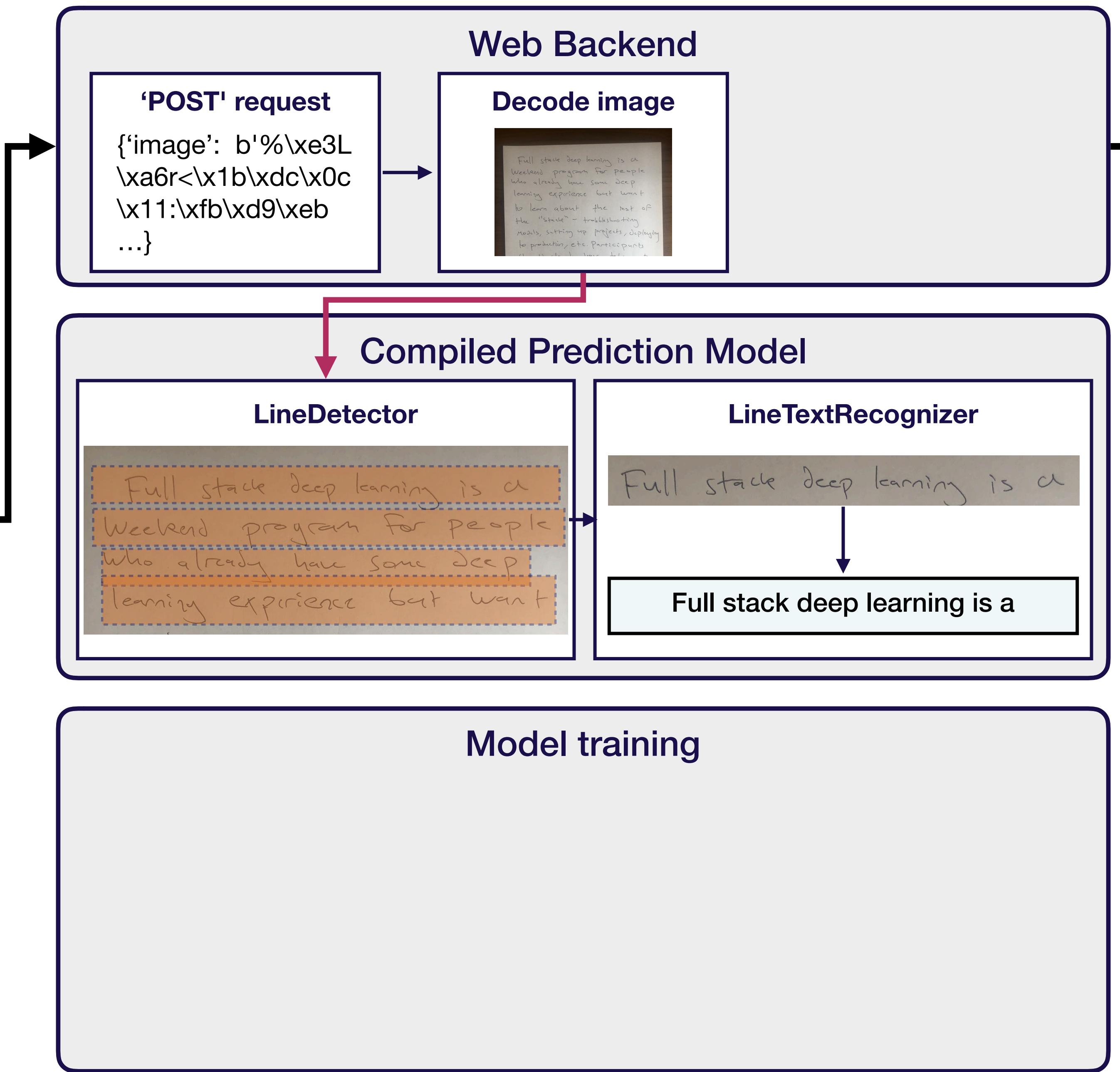
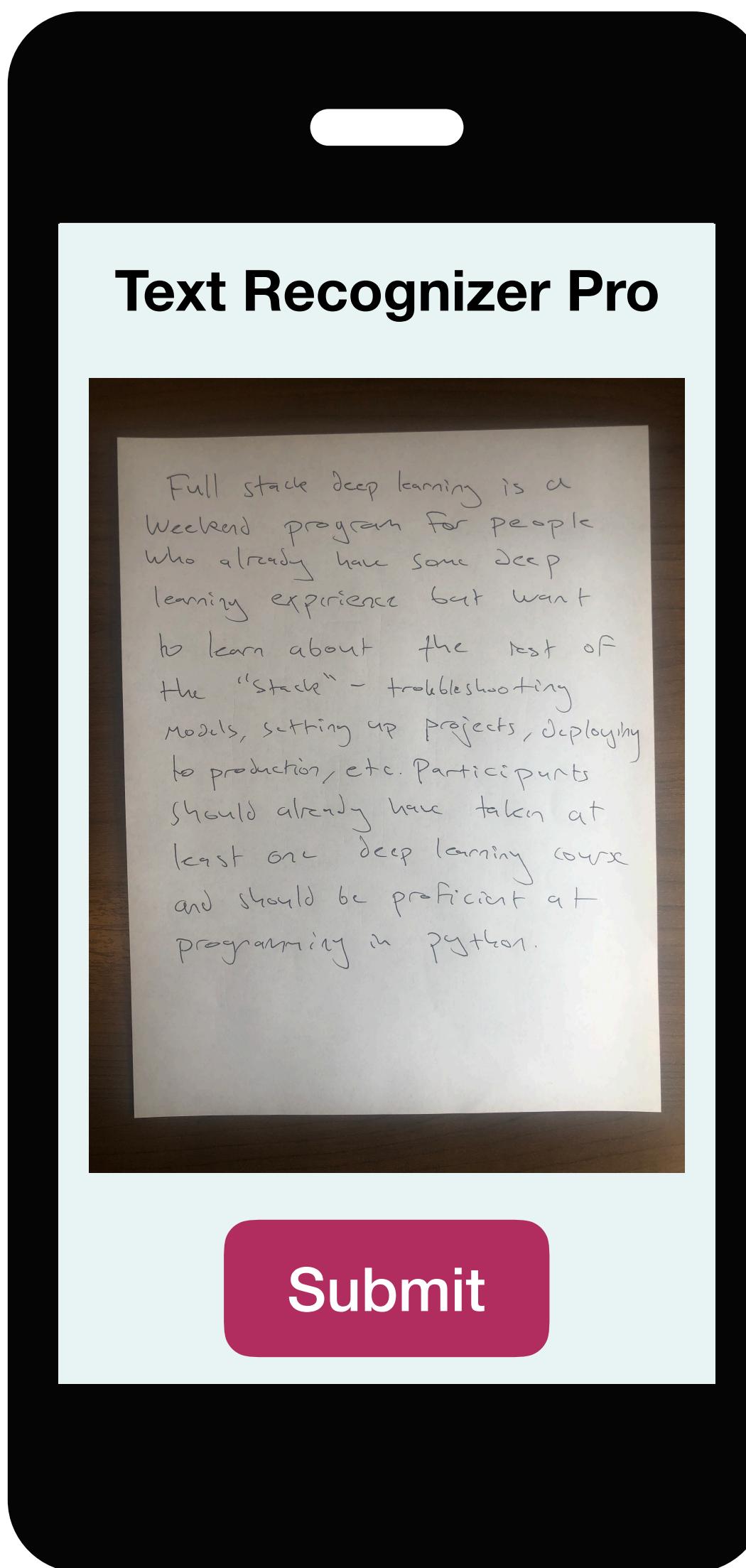
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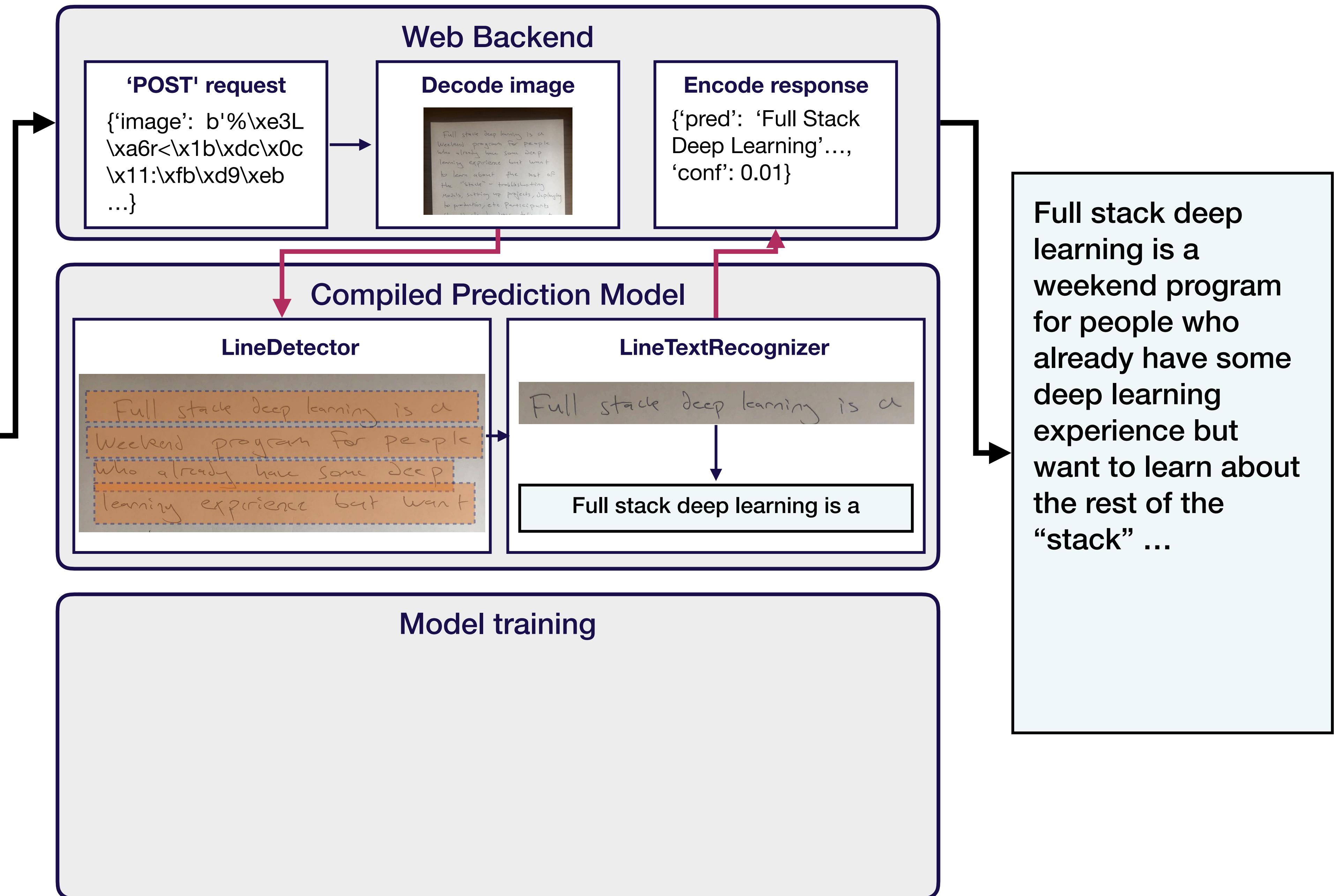
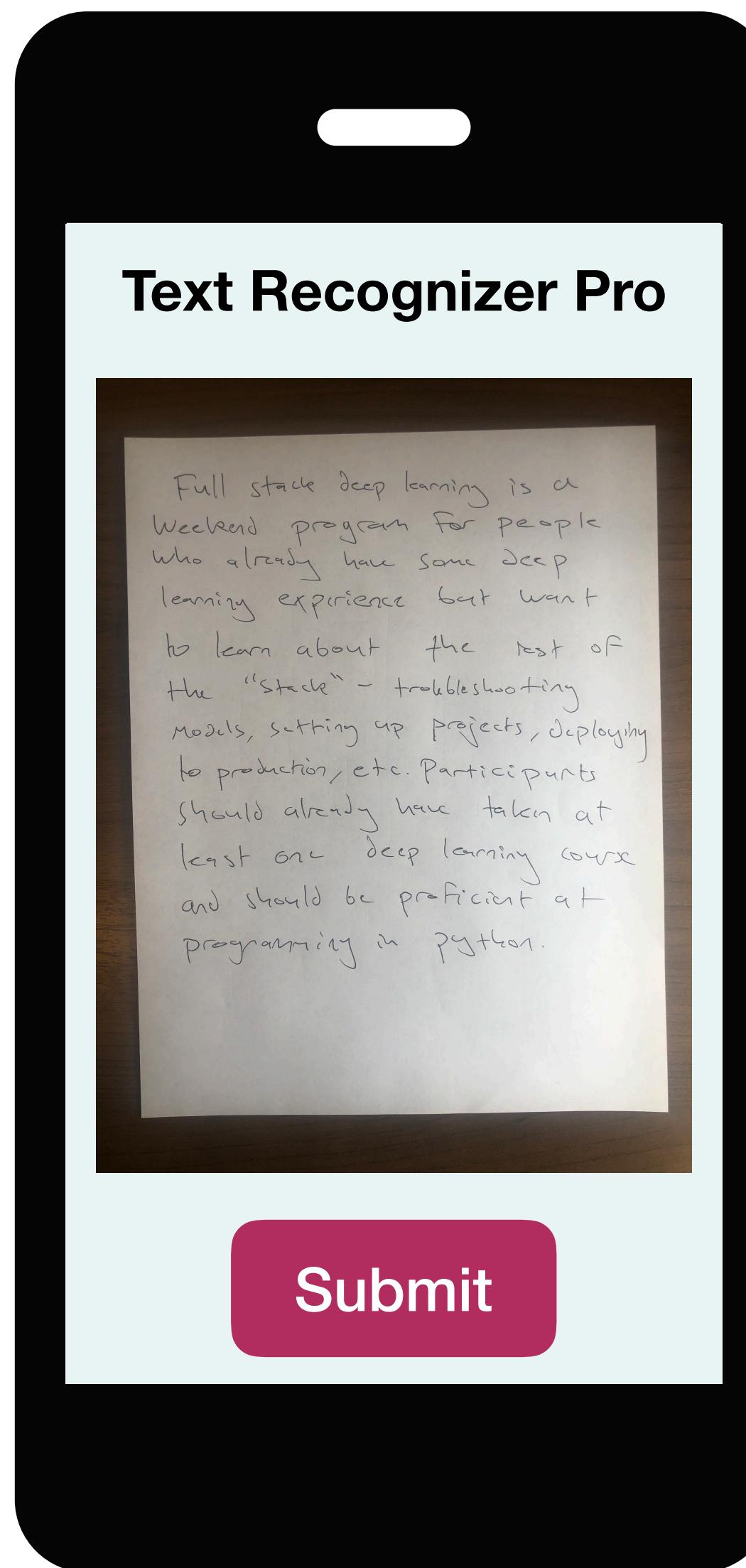
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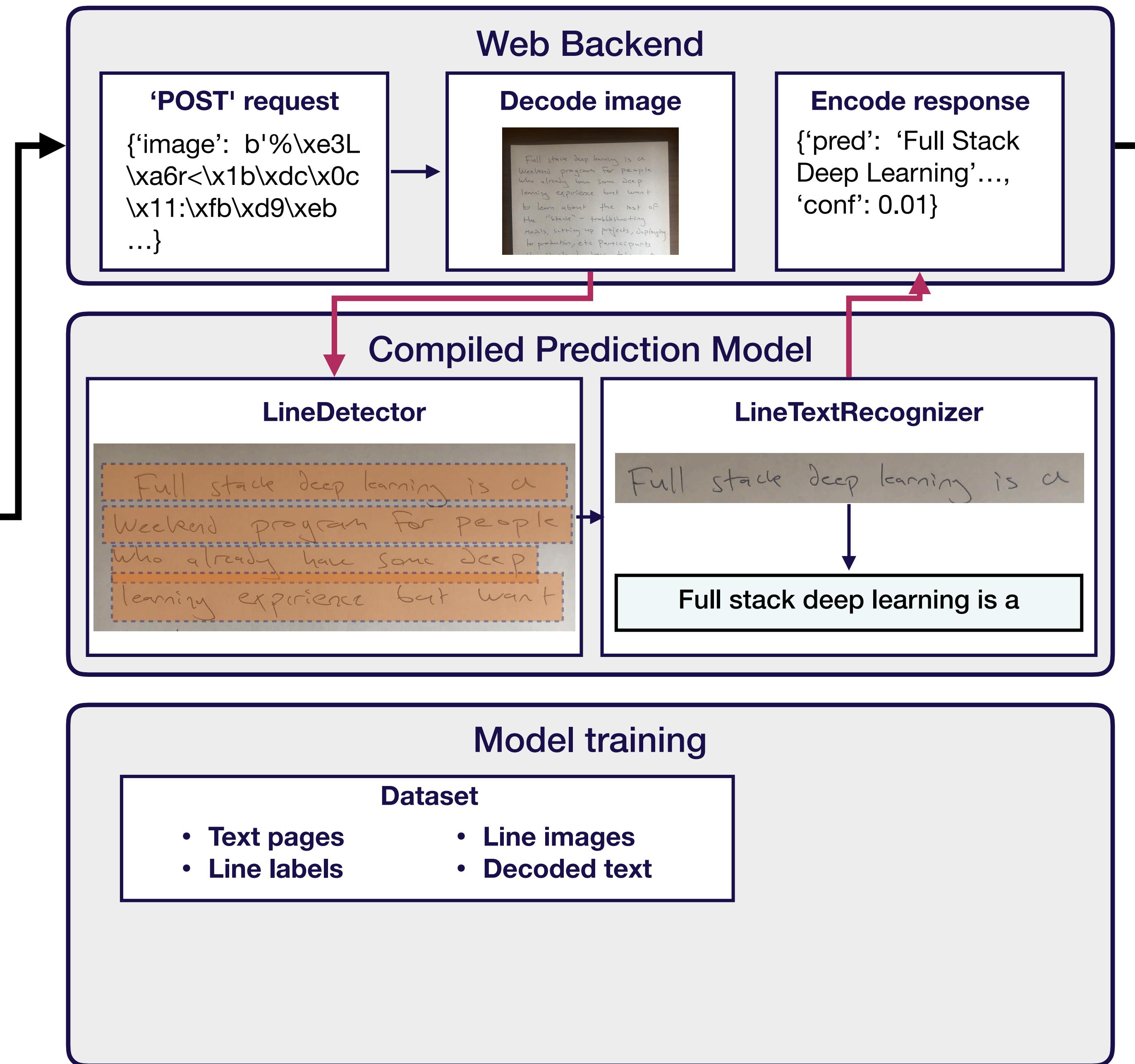
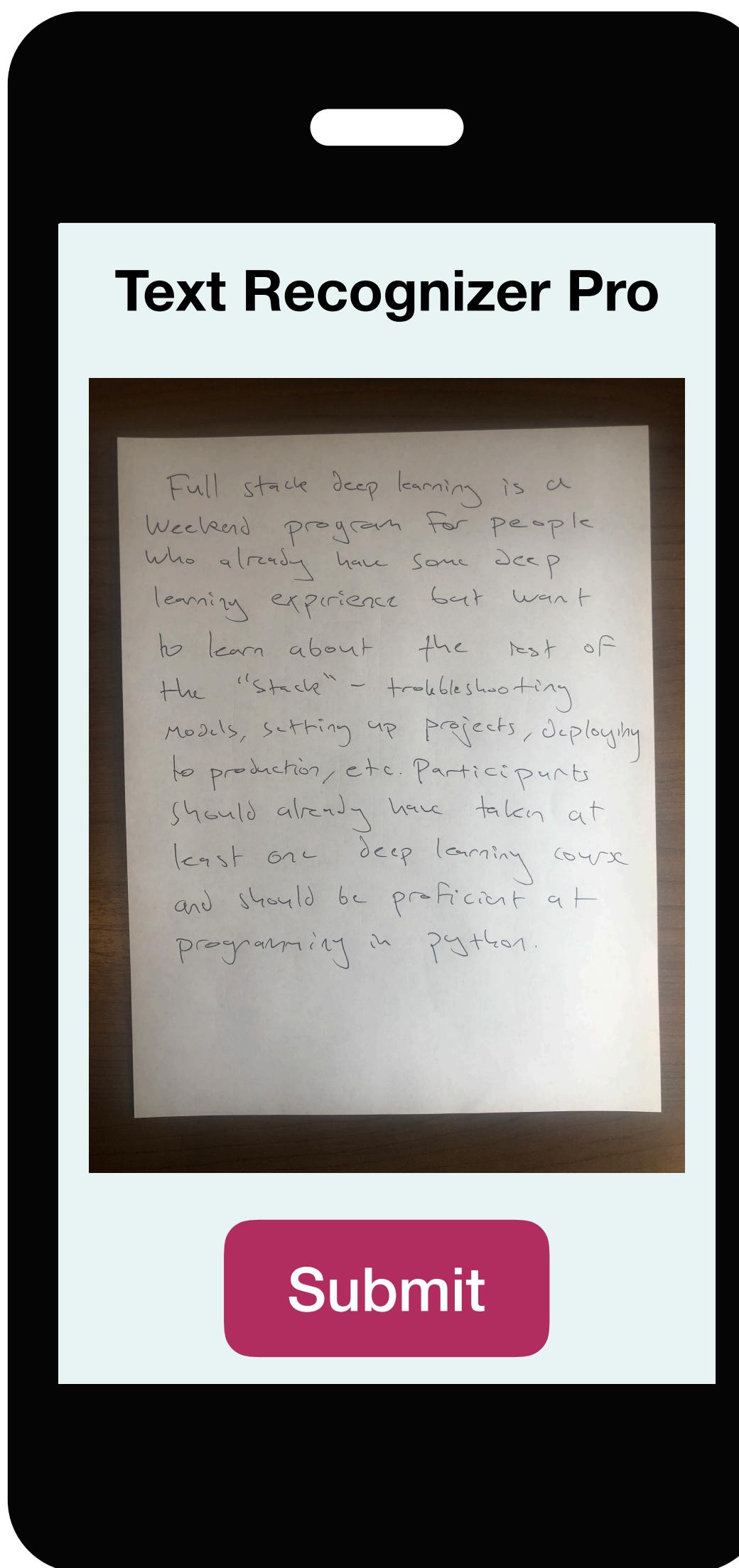


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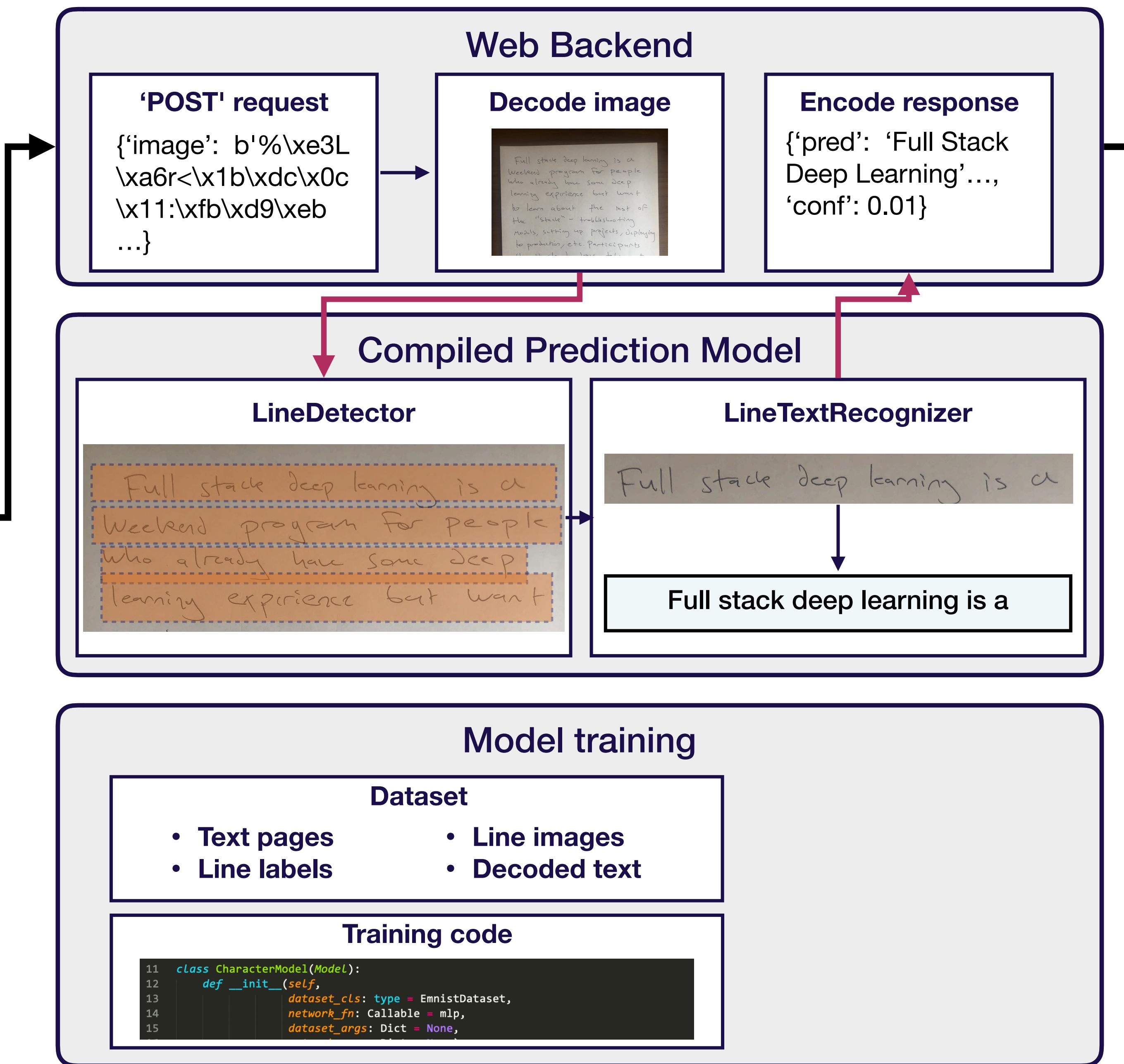
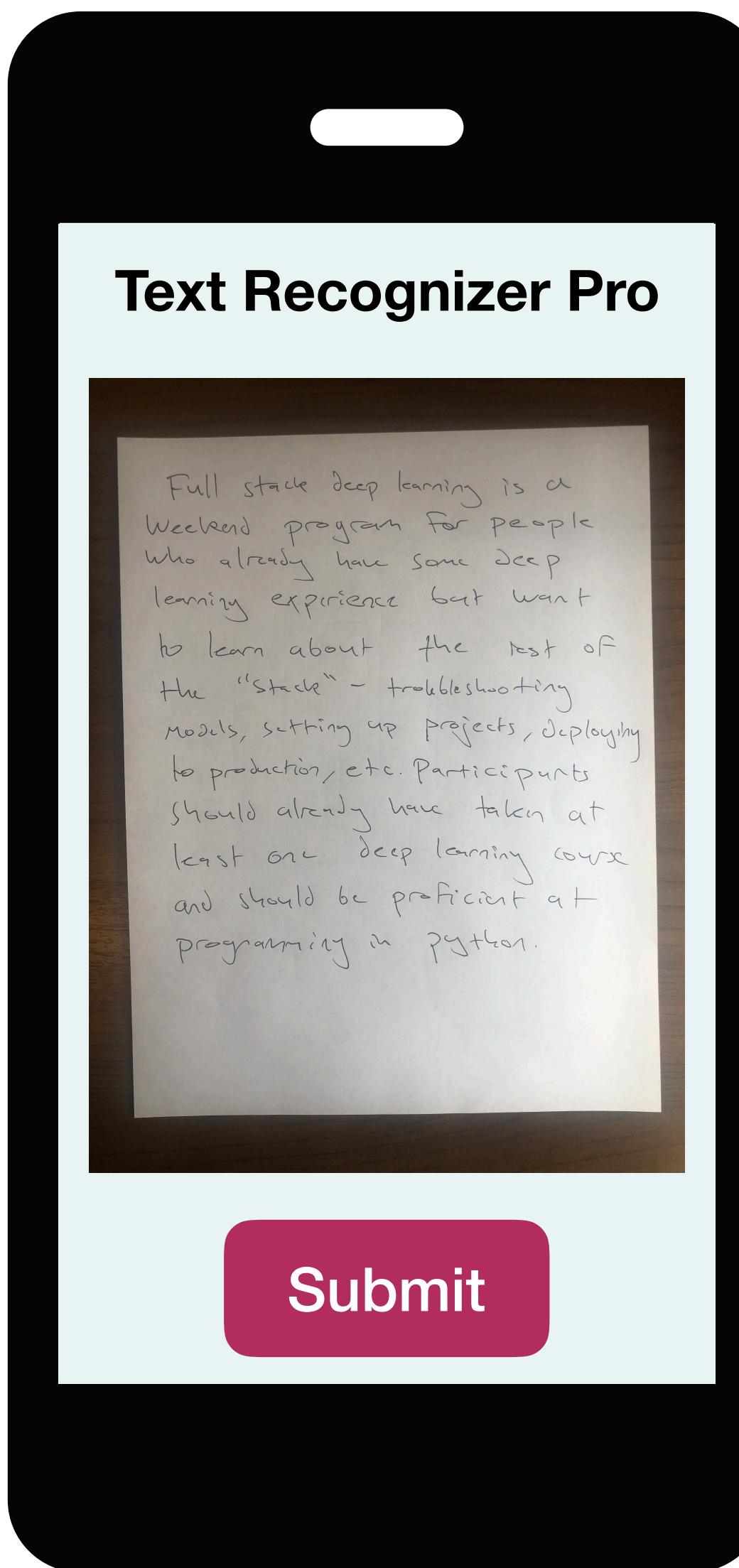


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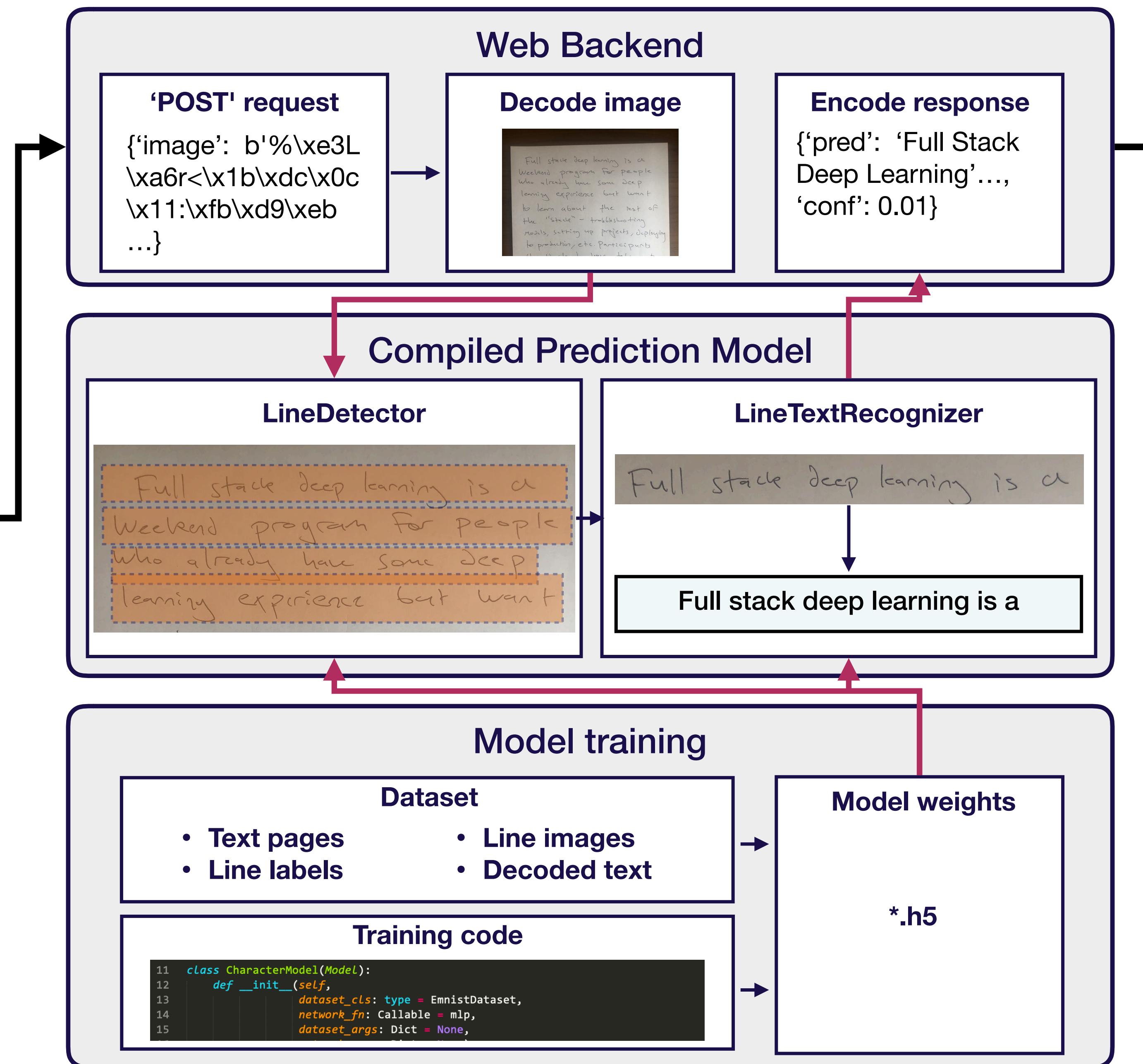
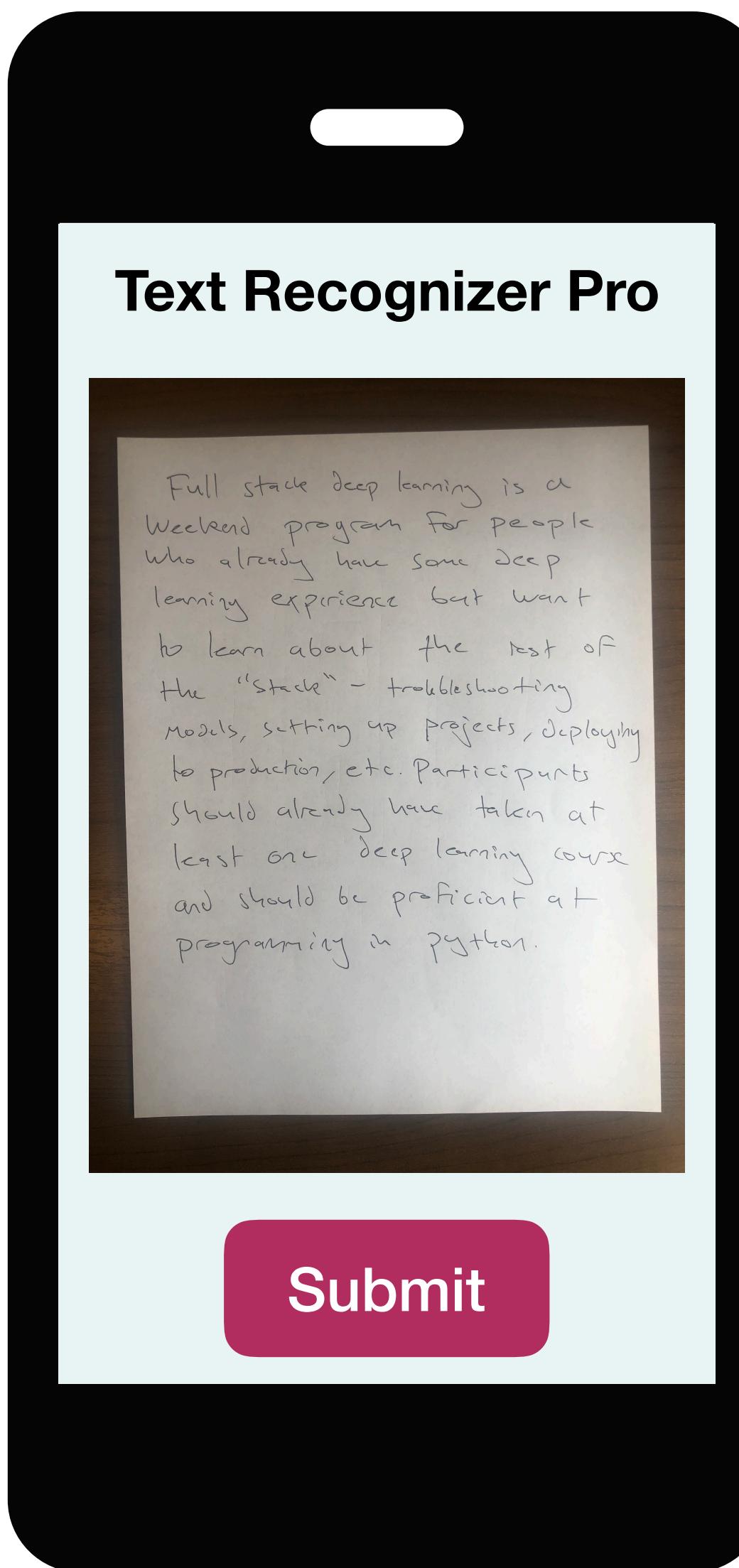




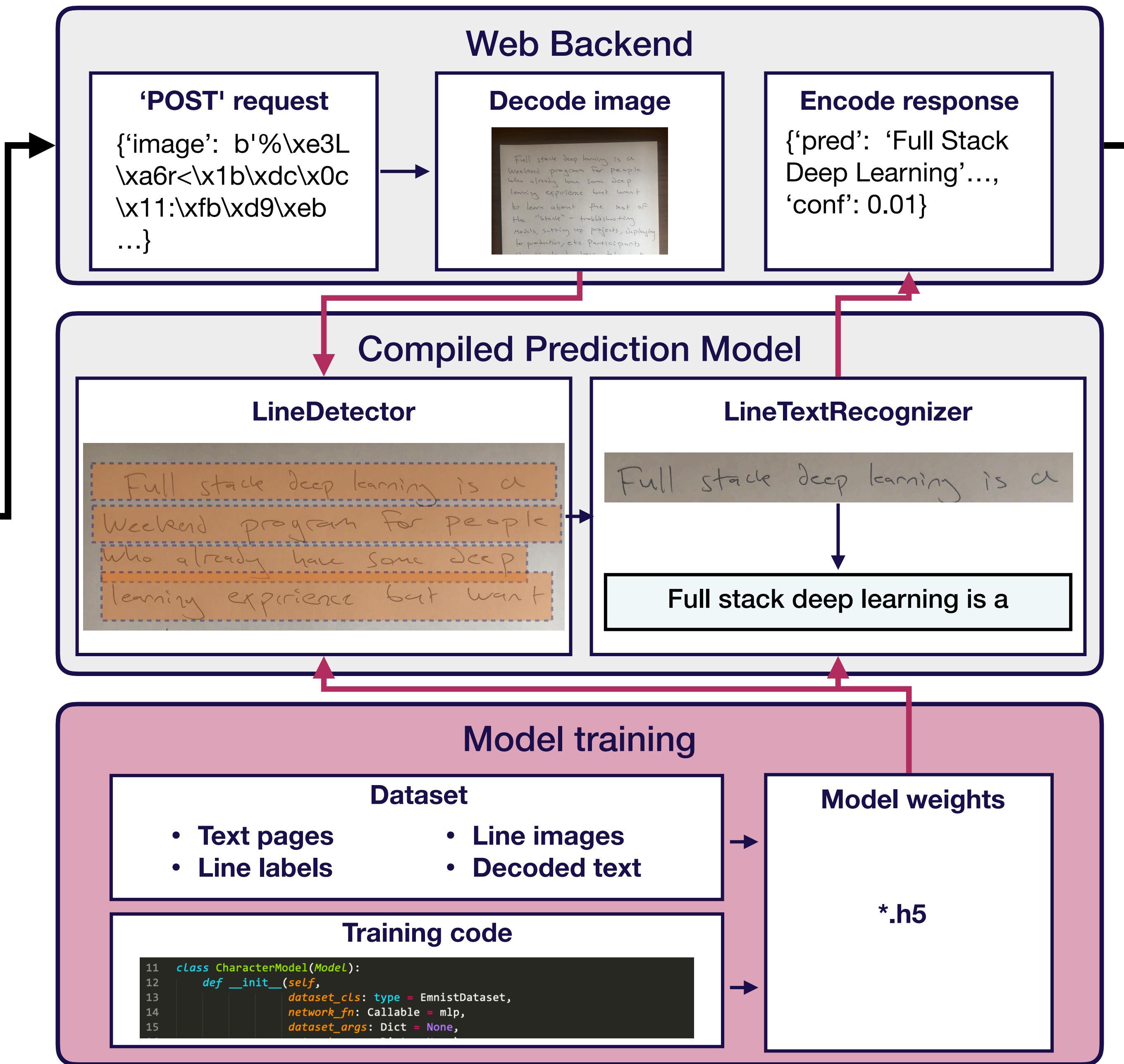
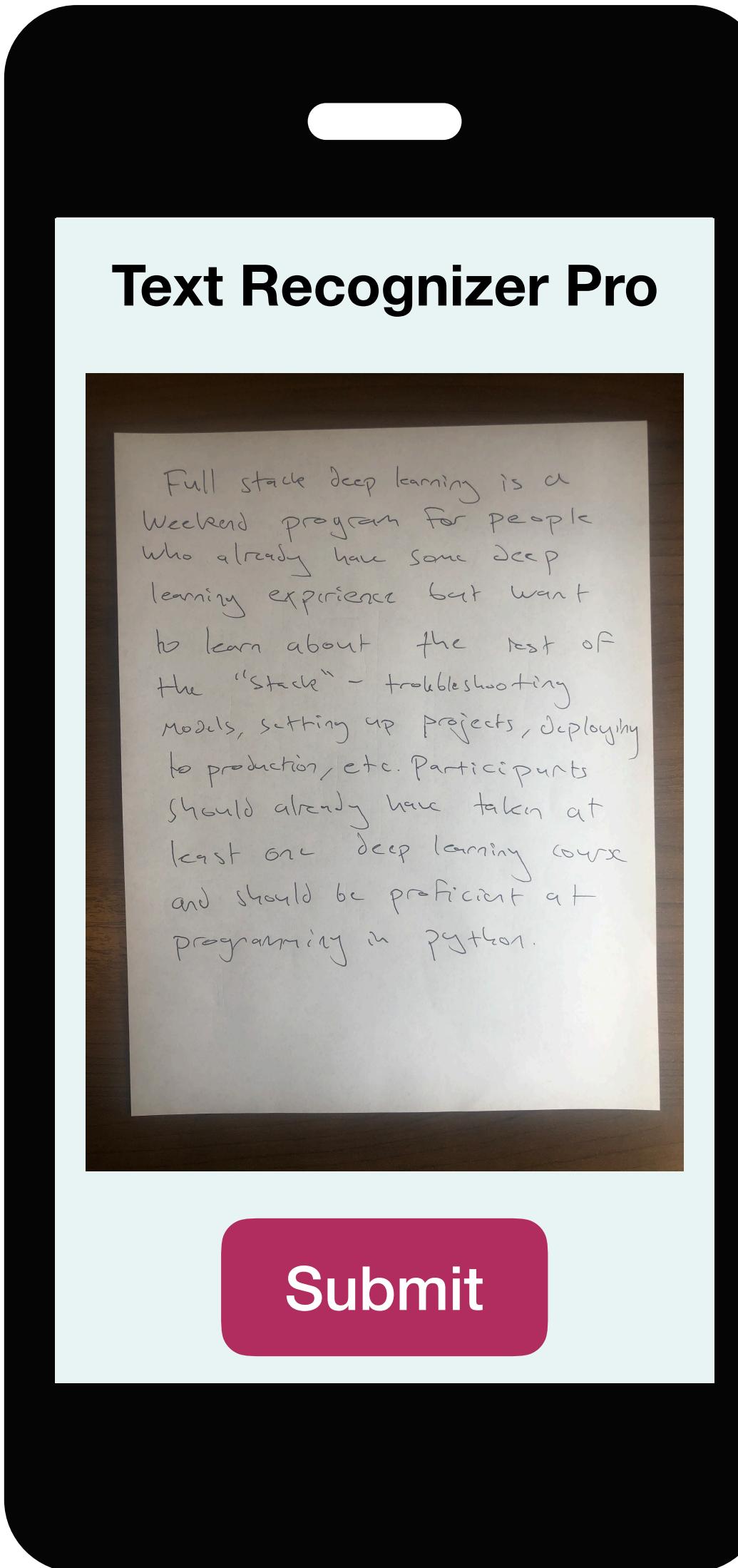
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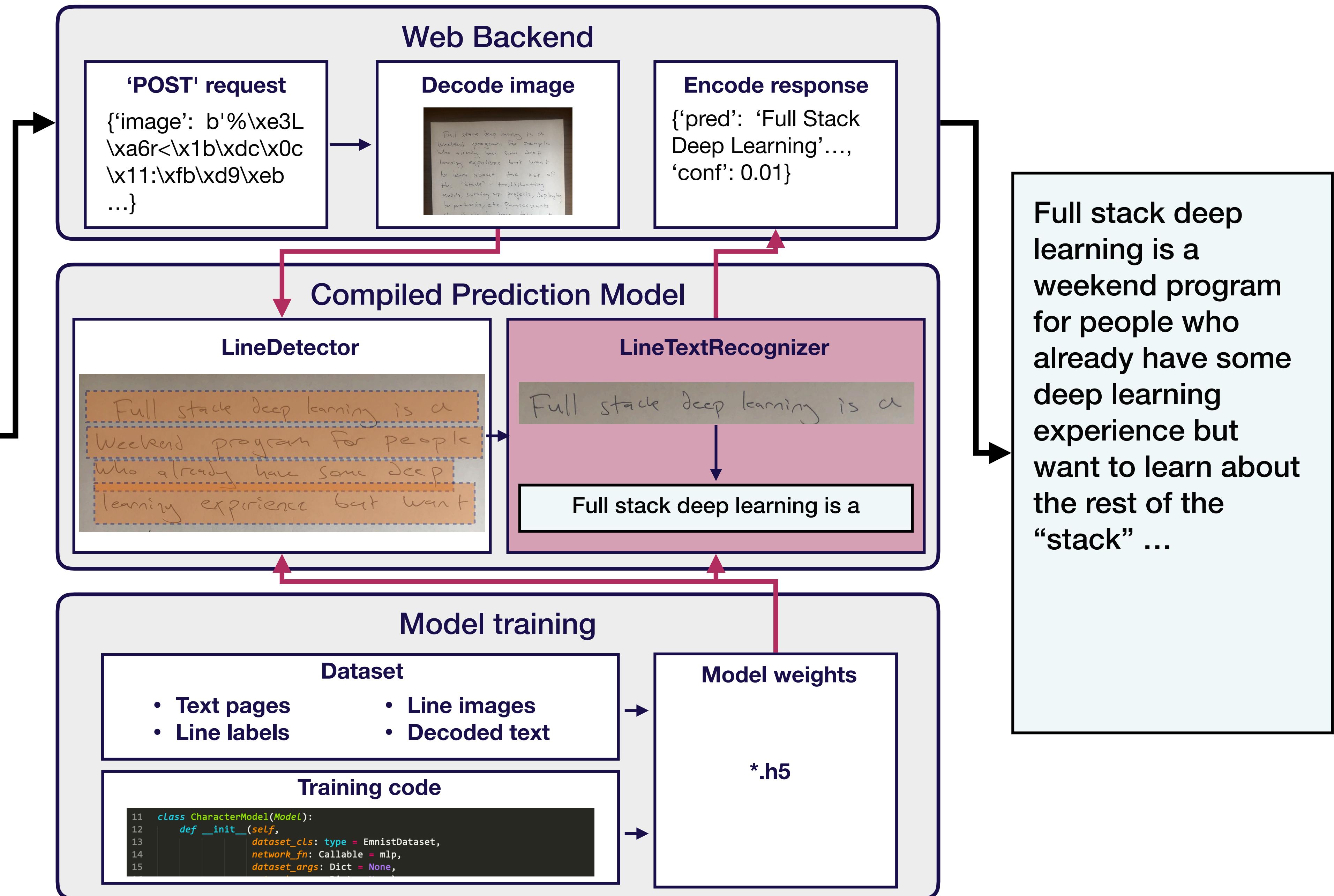
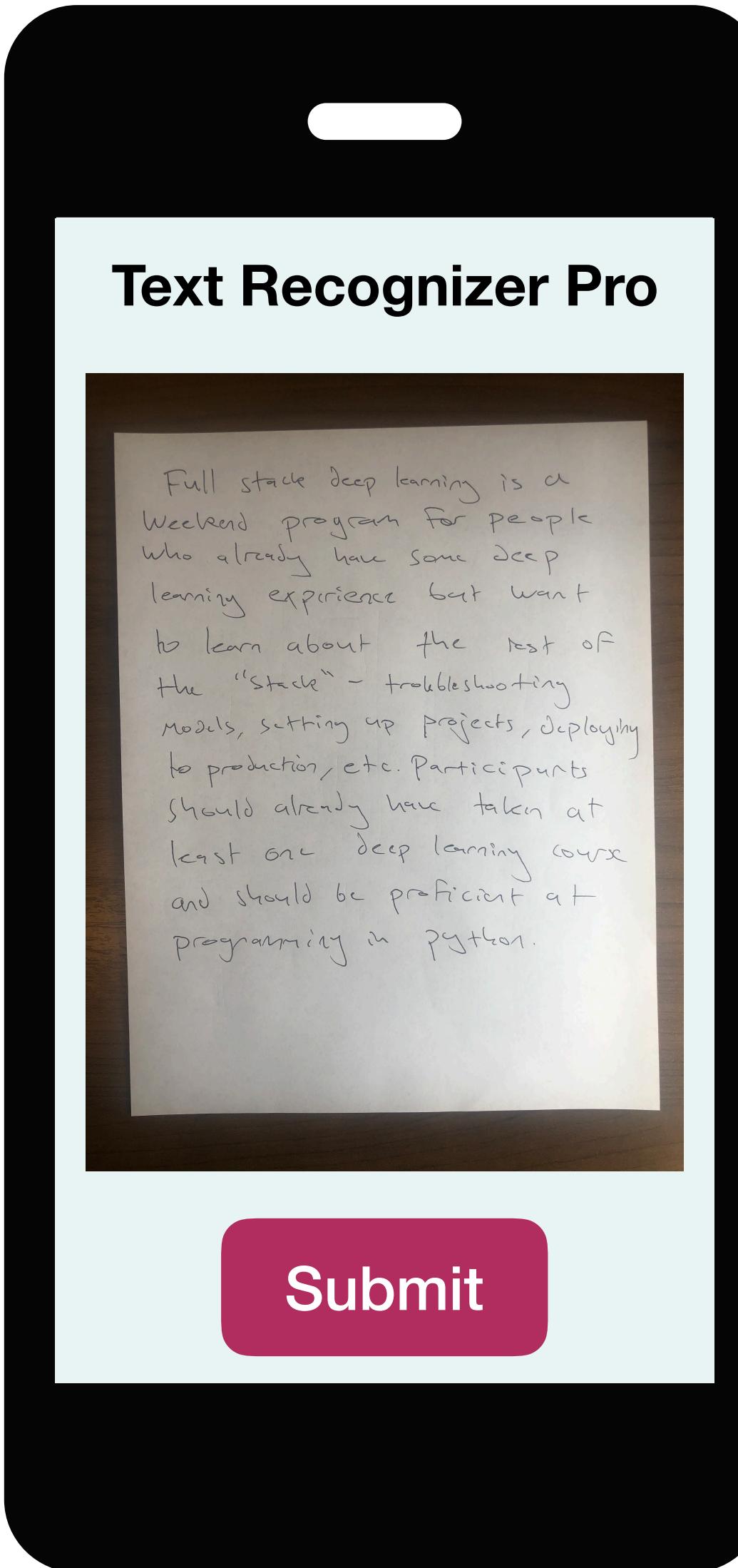
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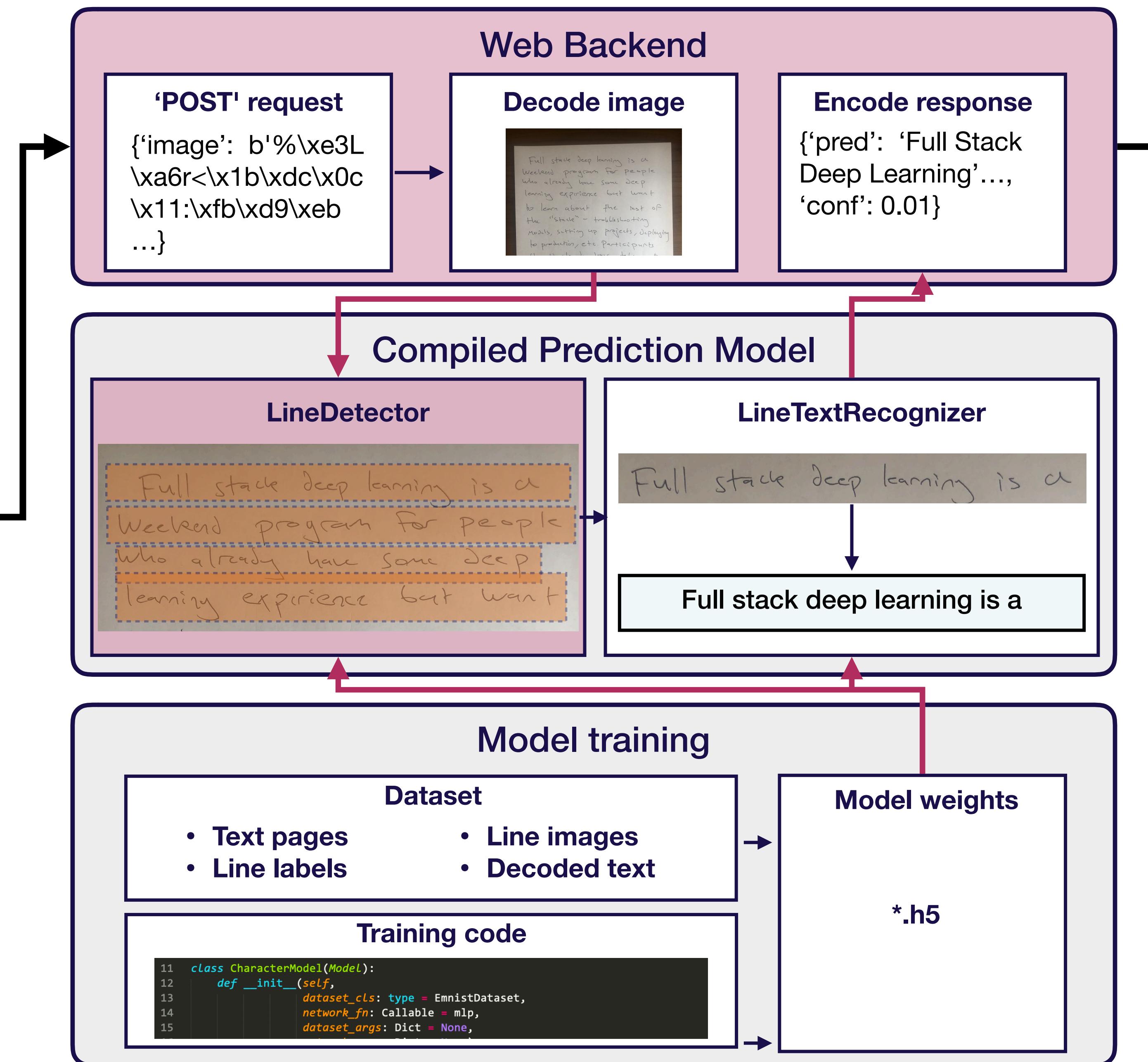
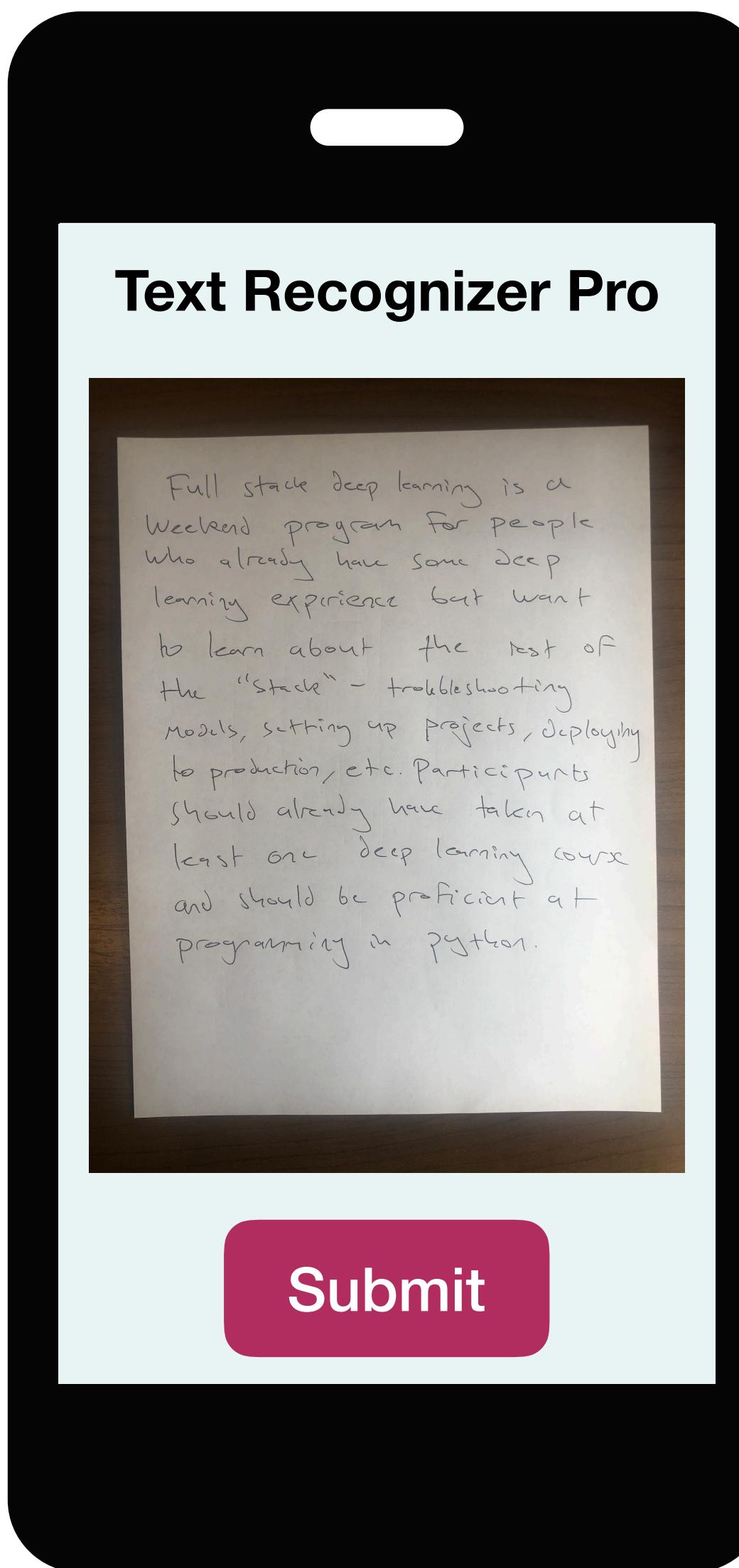


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Outline of the labs

- **Lab 1: Codebase walkthrough.**
- **Lab 2: Single-character prediction.** Before predicting full lines, try a simpler problem.
- **Lab 3: LineTextRecognizer.** Build a synthetic dataset. LSTM + CTC loss model.
- **Lab 4: Tools for experimentation.** Weights & Biases, parallel experiment running.
- **Lab 5. Experimentation.** Try some things and run some things overnight.
- **Lab 6. LineDetector.** Train the line detection model.
- **Lab 7. Data.** Managing data, including label and versioning.
- **Lab 8. Continuous integration.** Testing your model.
- **Lab 9. Deployment.** Put the model into production.

Go here

<https://github.com/full-stack-deep-learning/fsdl-text-recognizer-project>

W&B Access Code

fsdl-deep-cajon

LineTextRecognizer Model architecture

Outputs

the quick brown fox

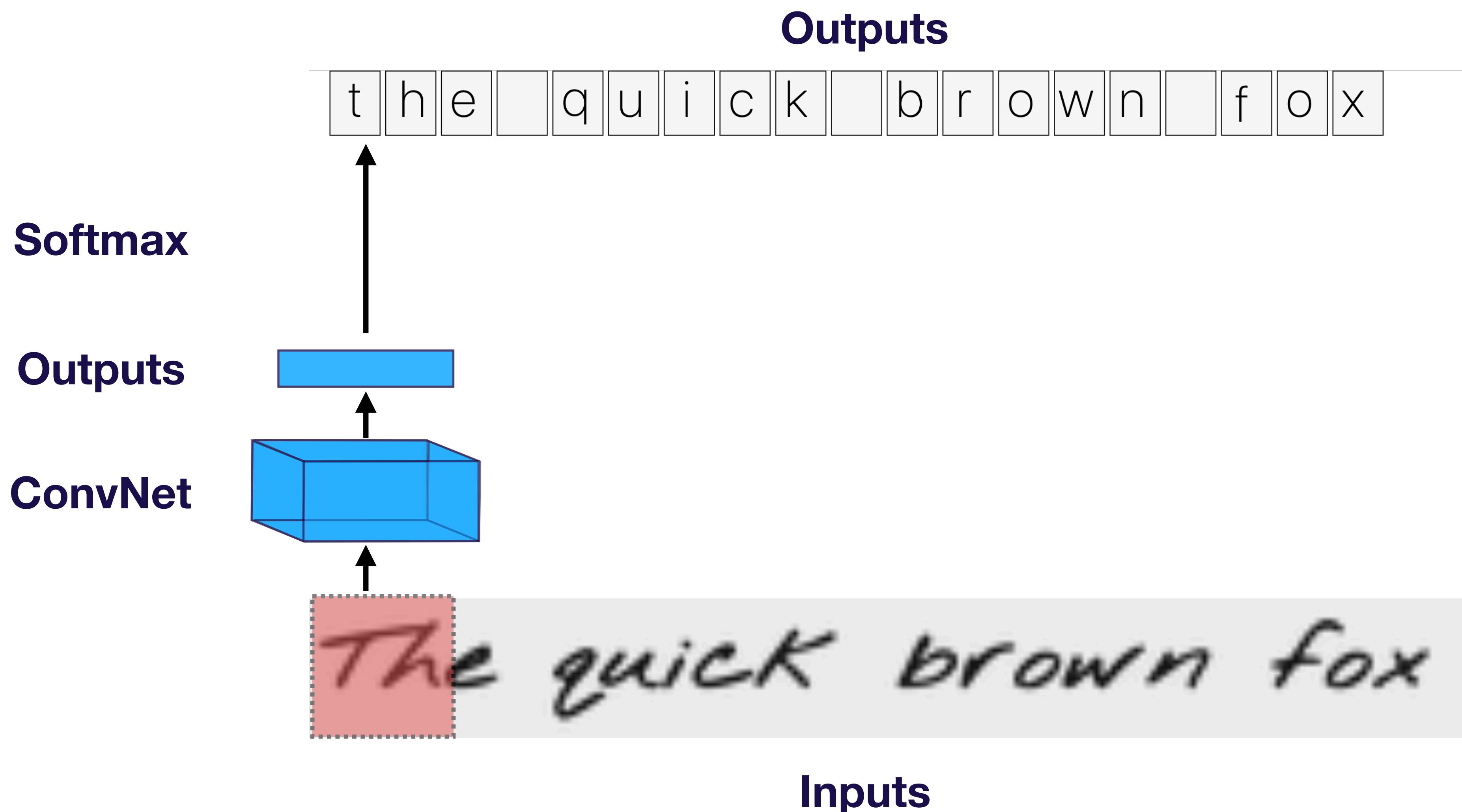


The quick brown fox

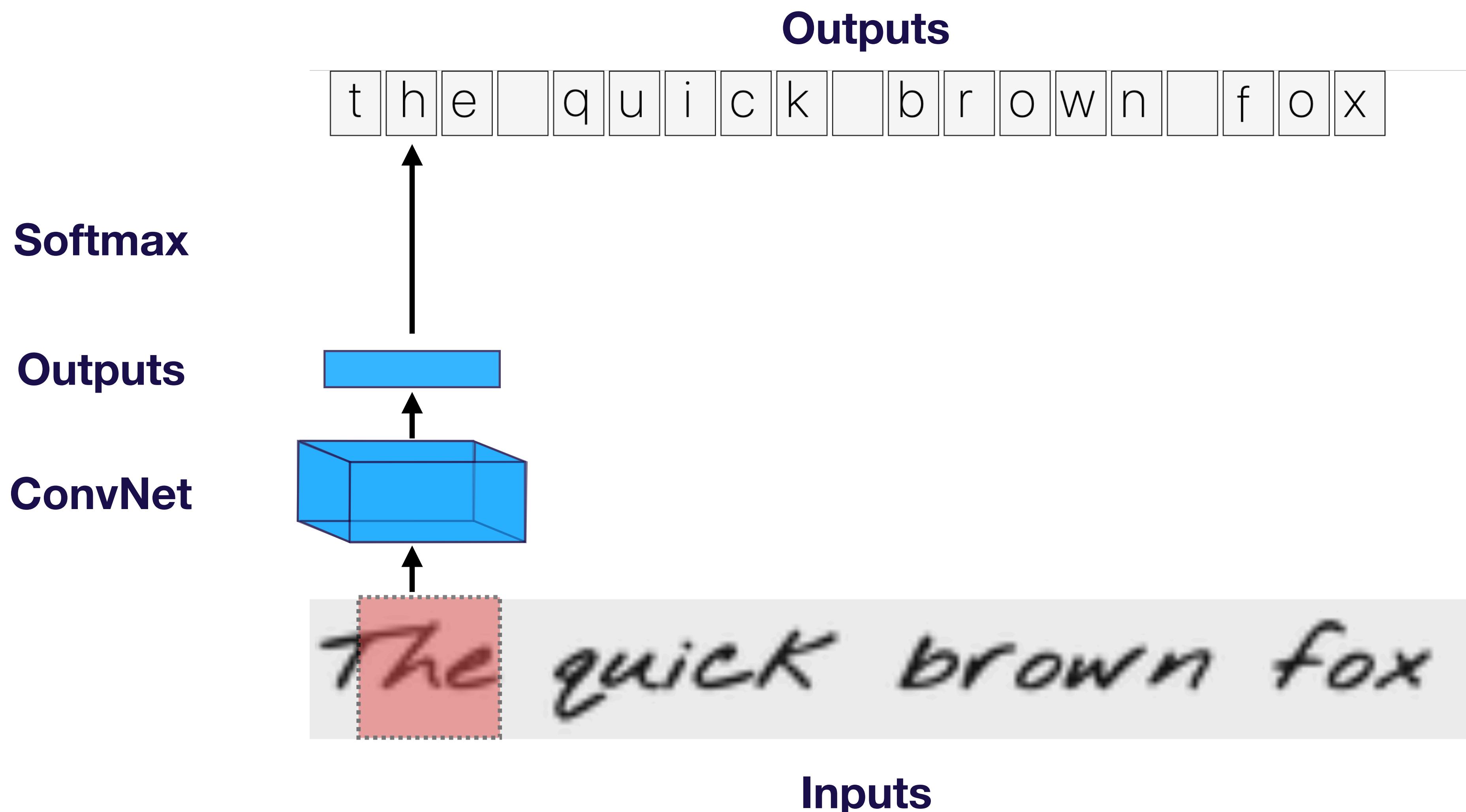
Inputs

Sequence Modeling with CTC, Awni Hannun, Distill 10.23915/distill.00008

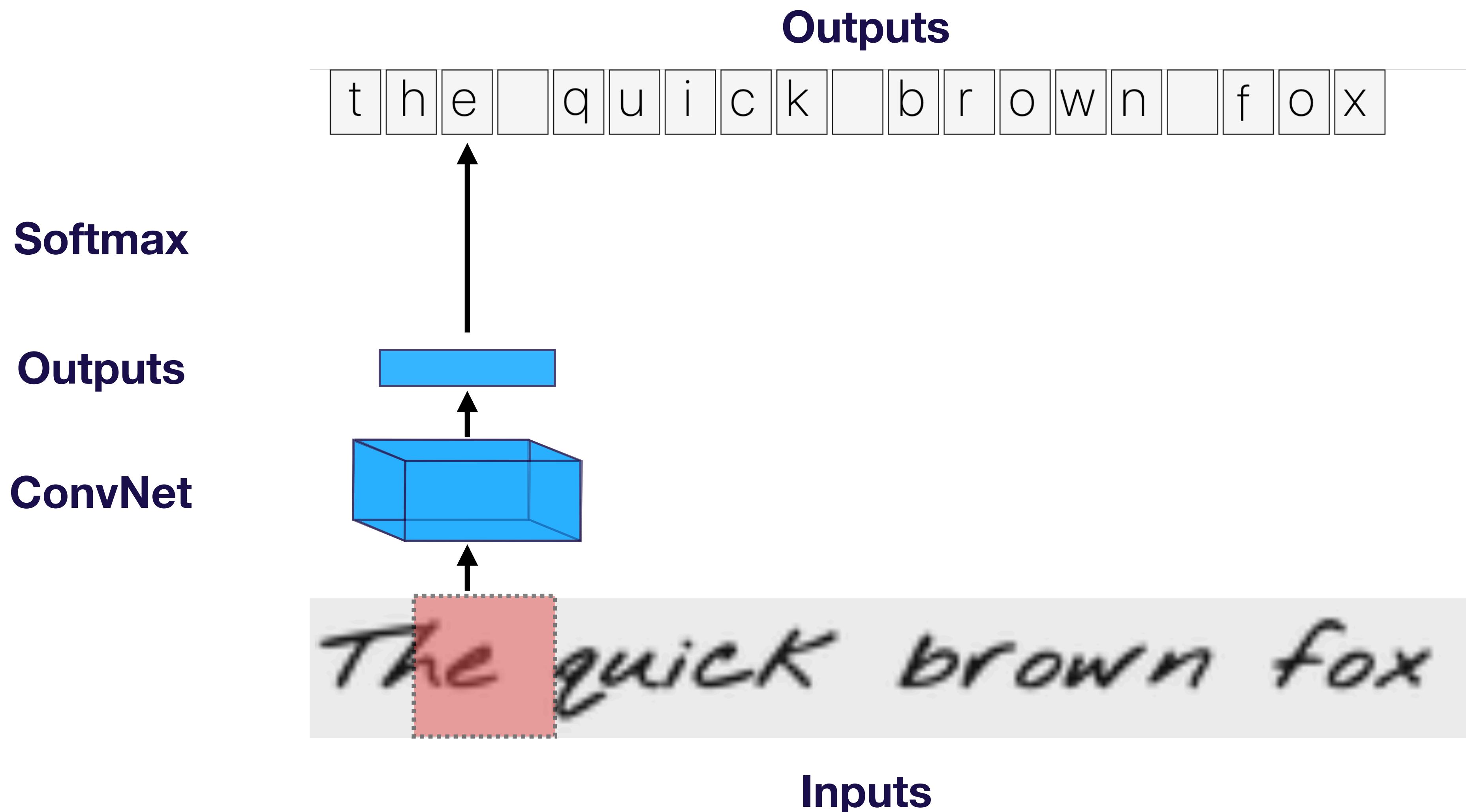
A baseline model architecture



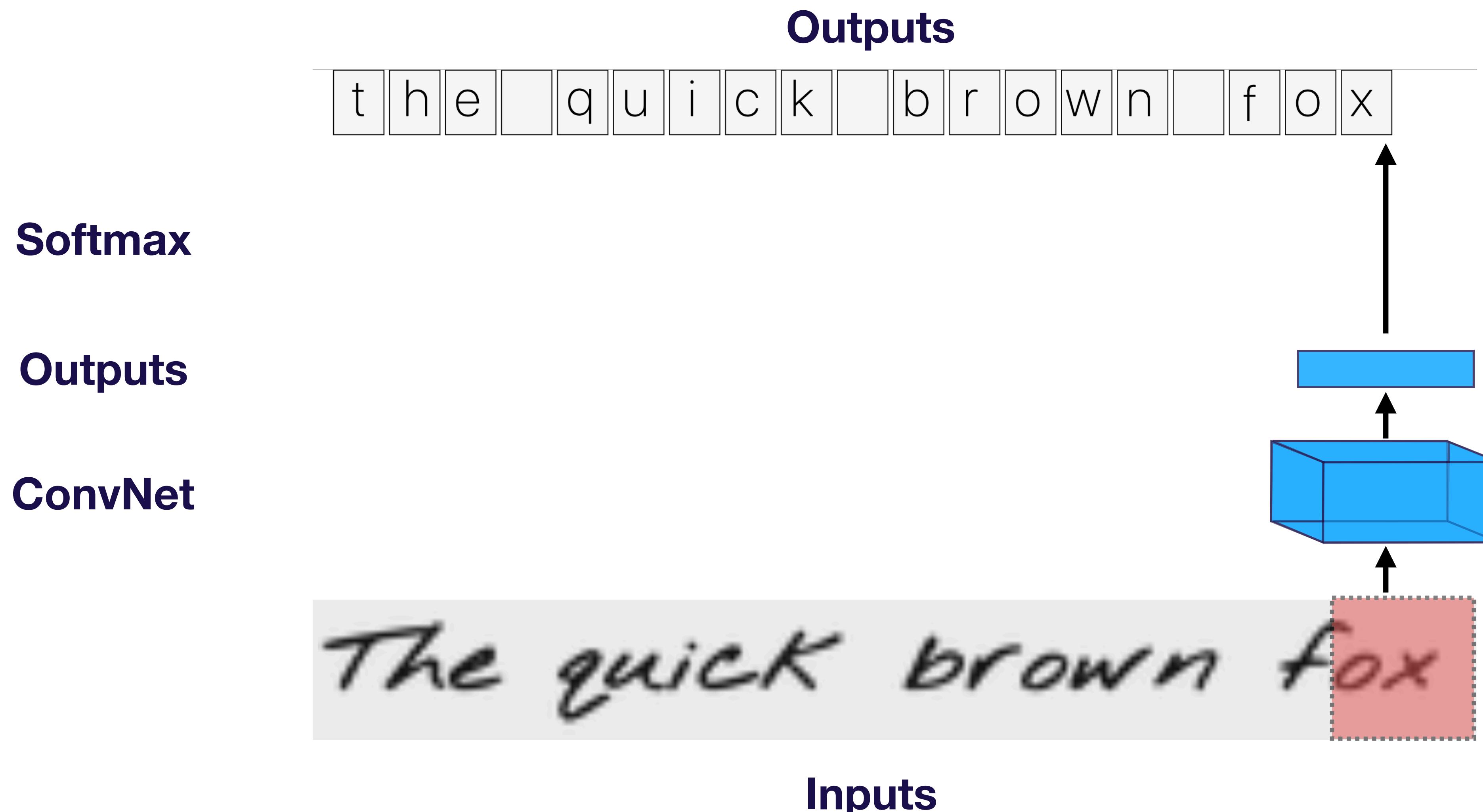
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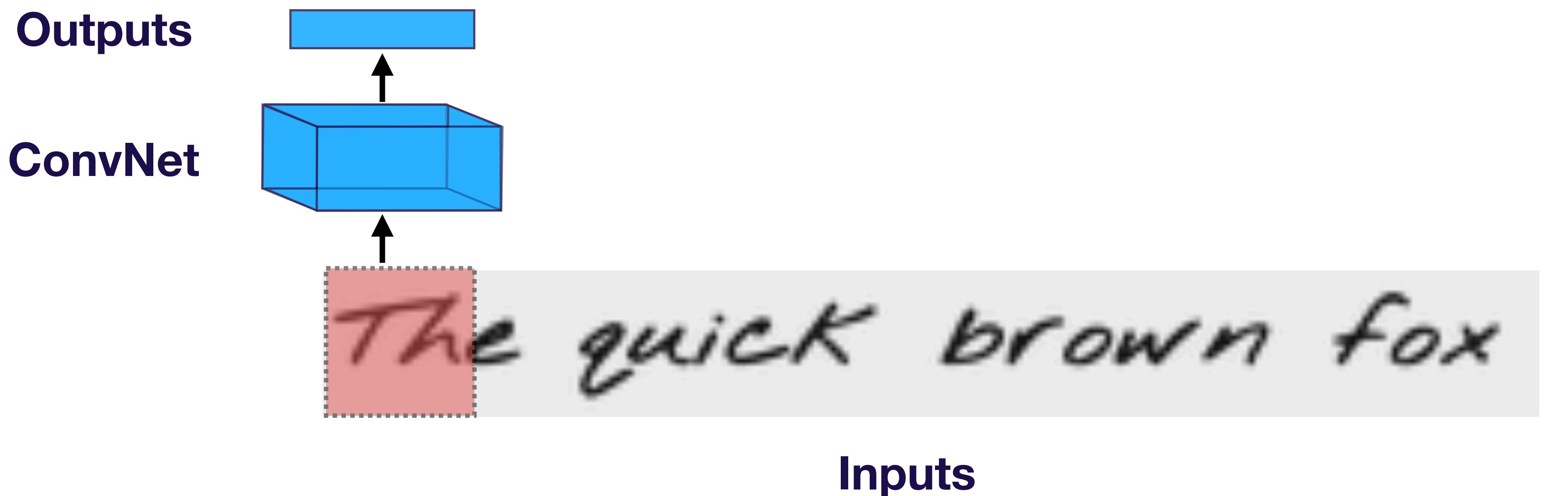


Let's give the model some context

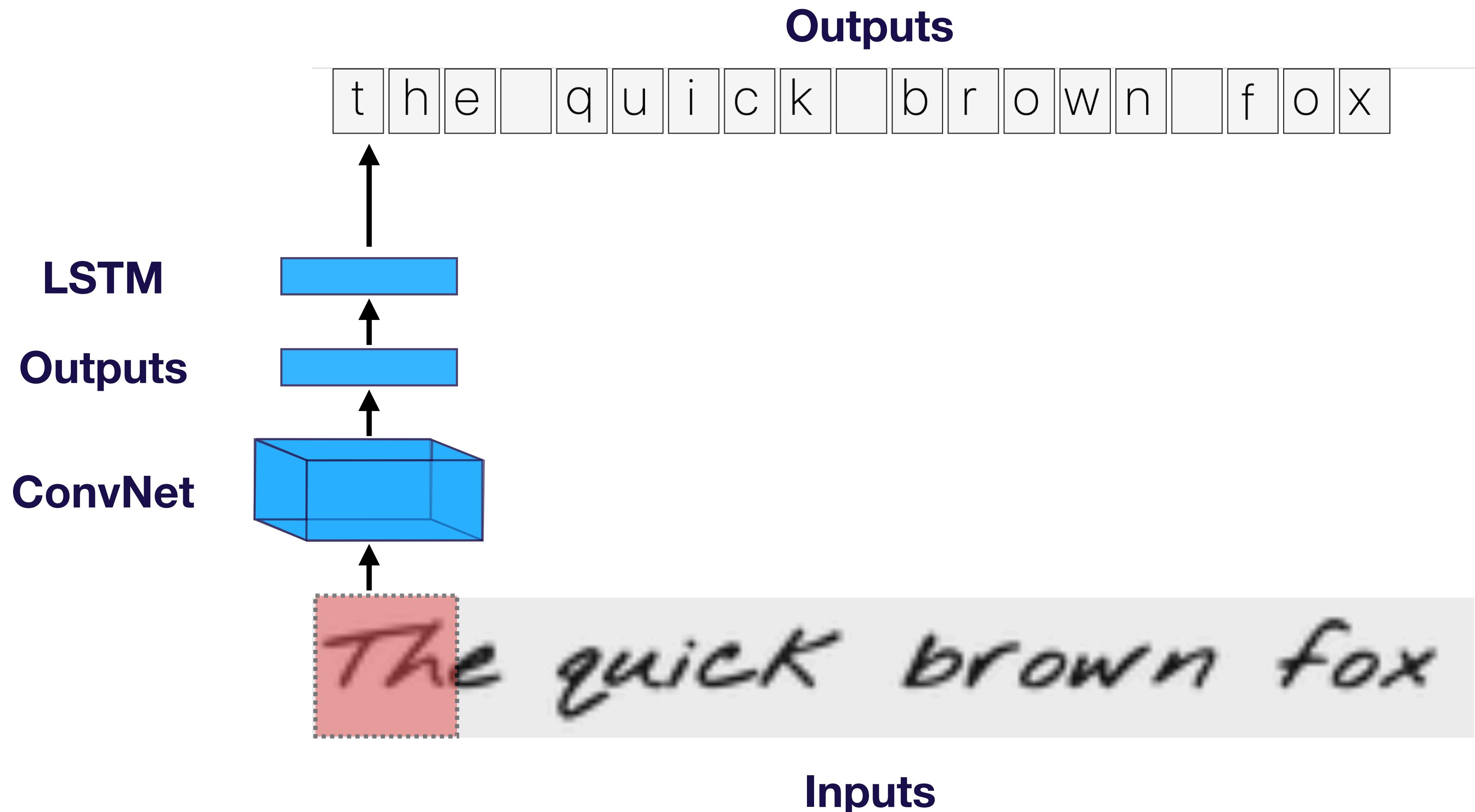
Outputs

the quick brown fox

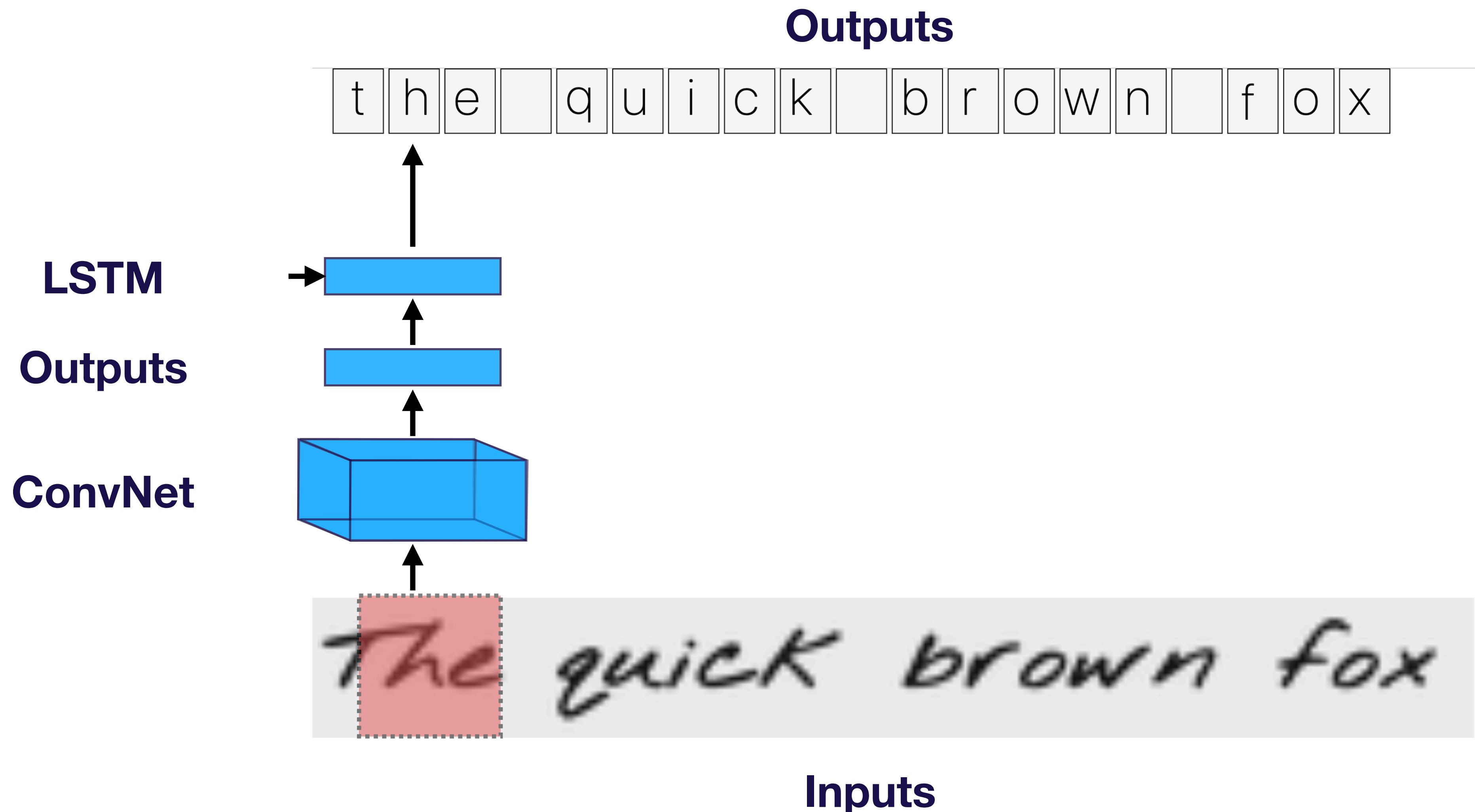
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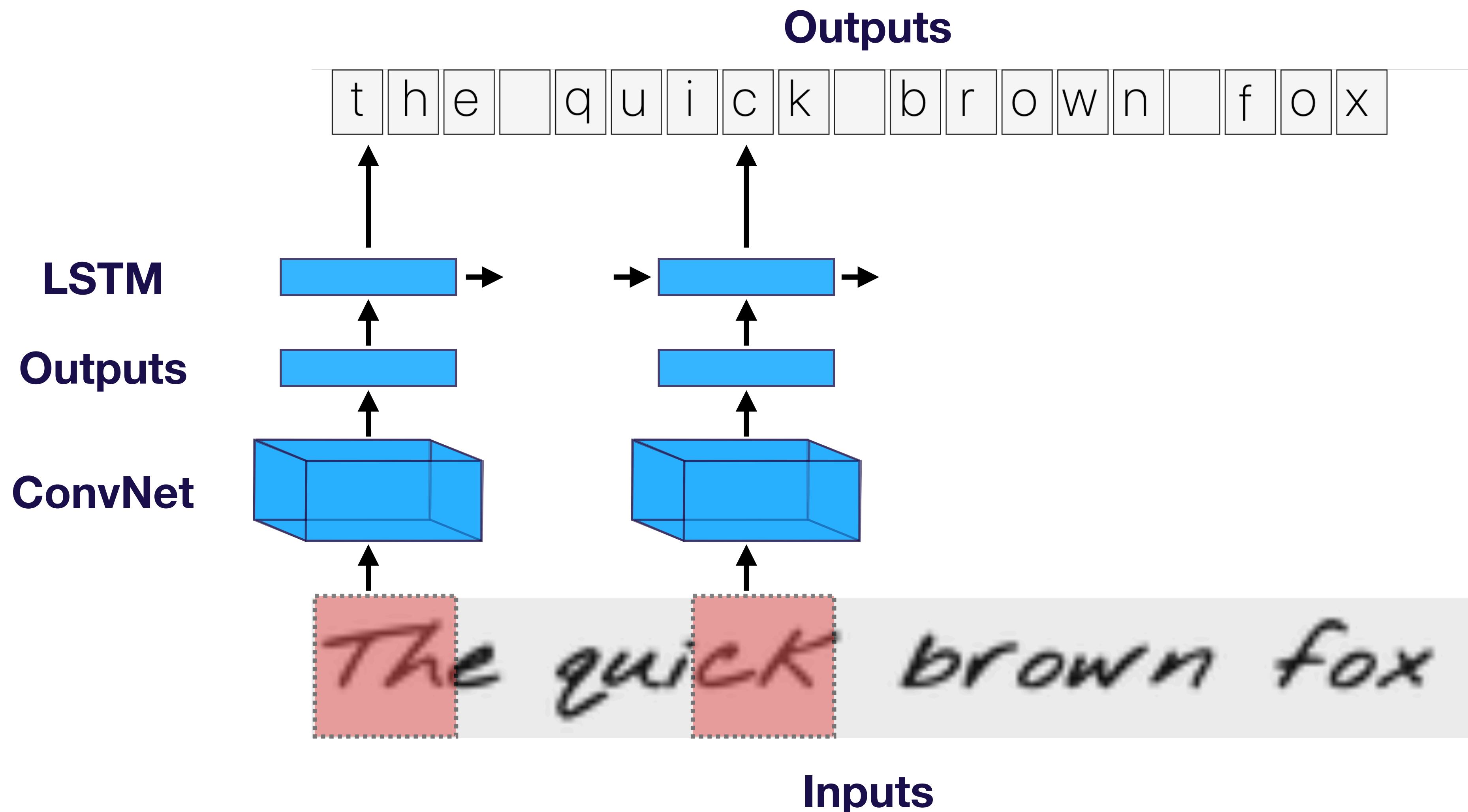
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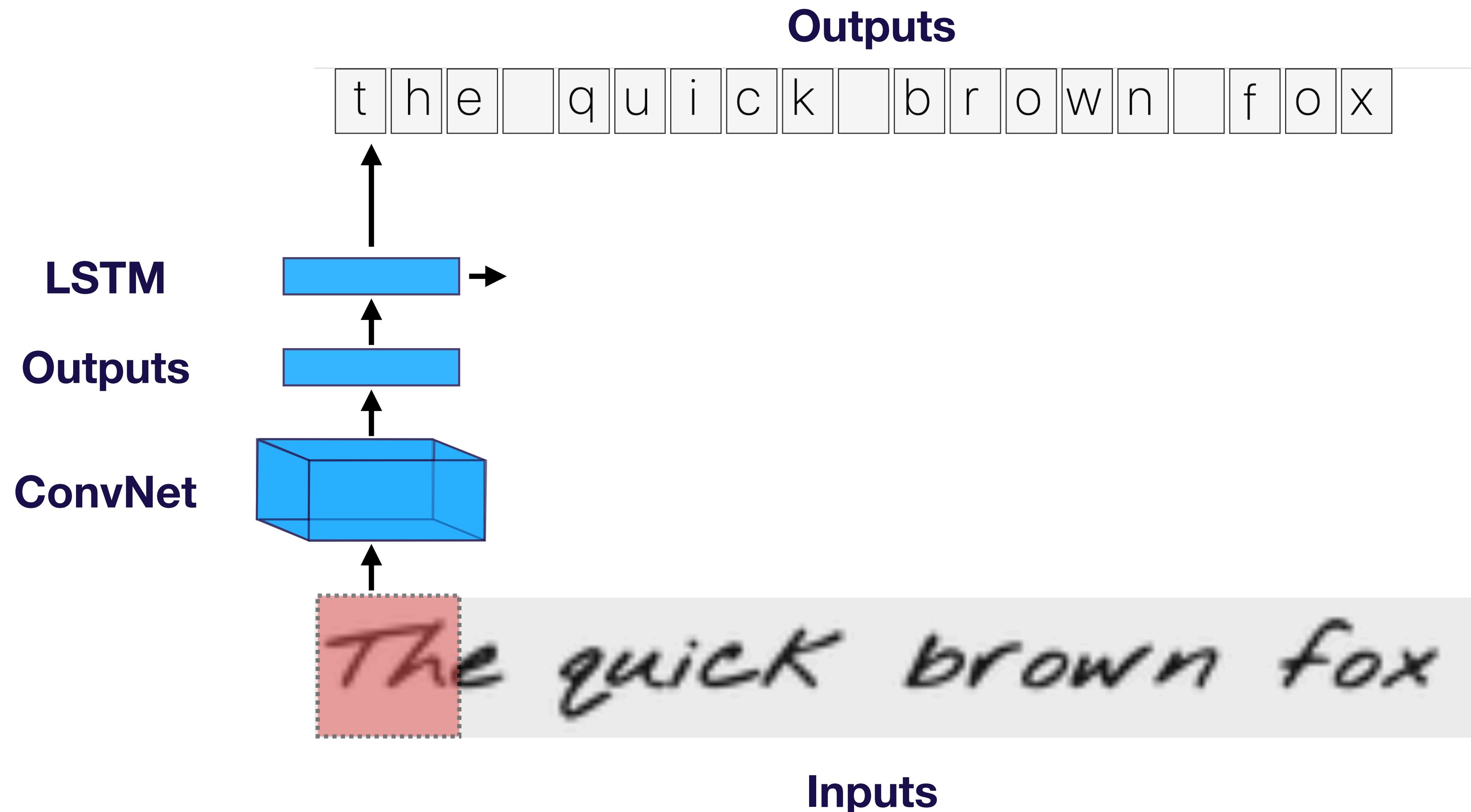
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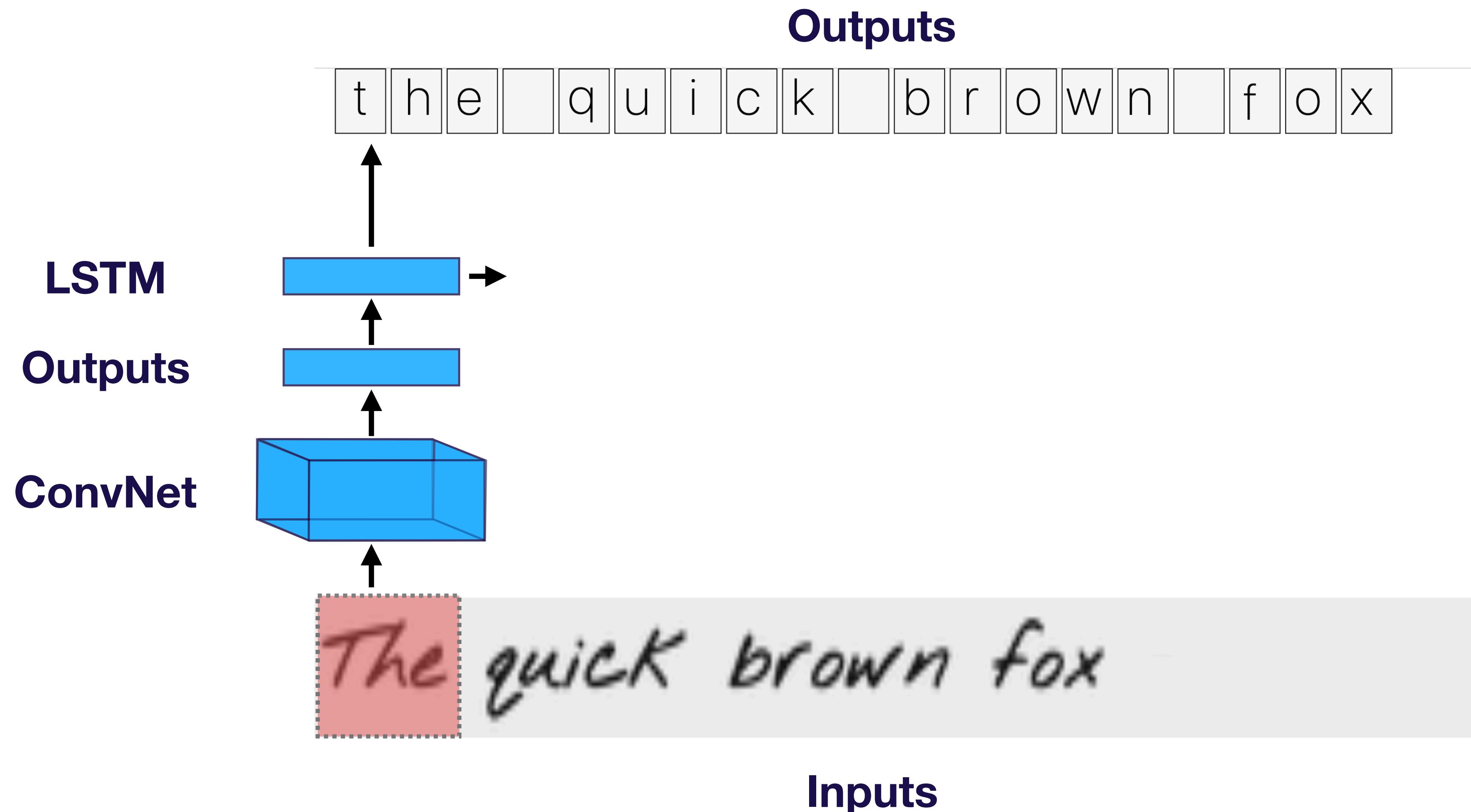
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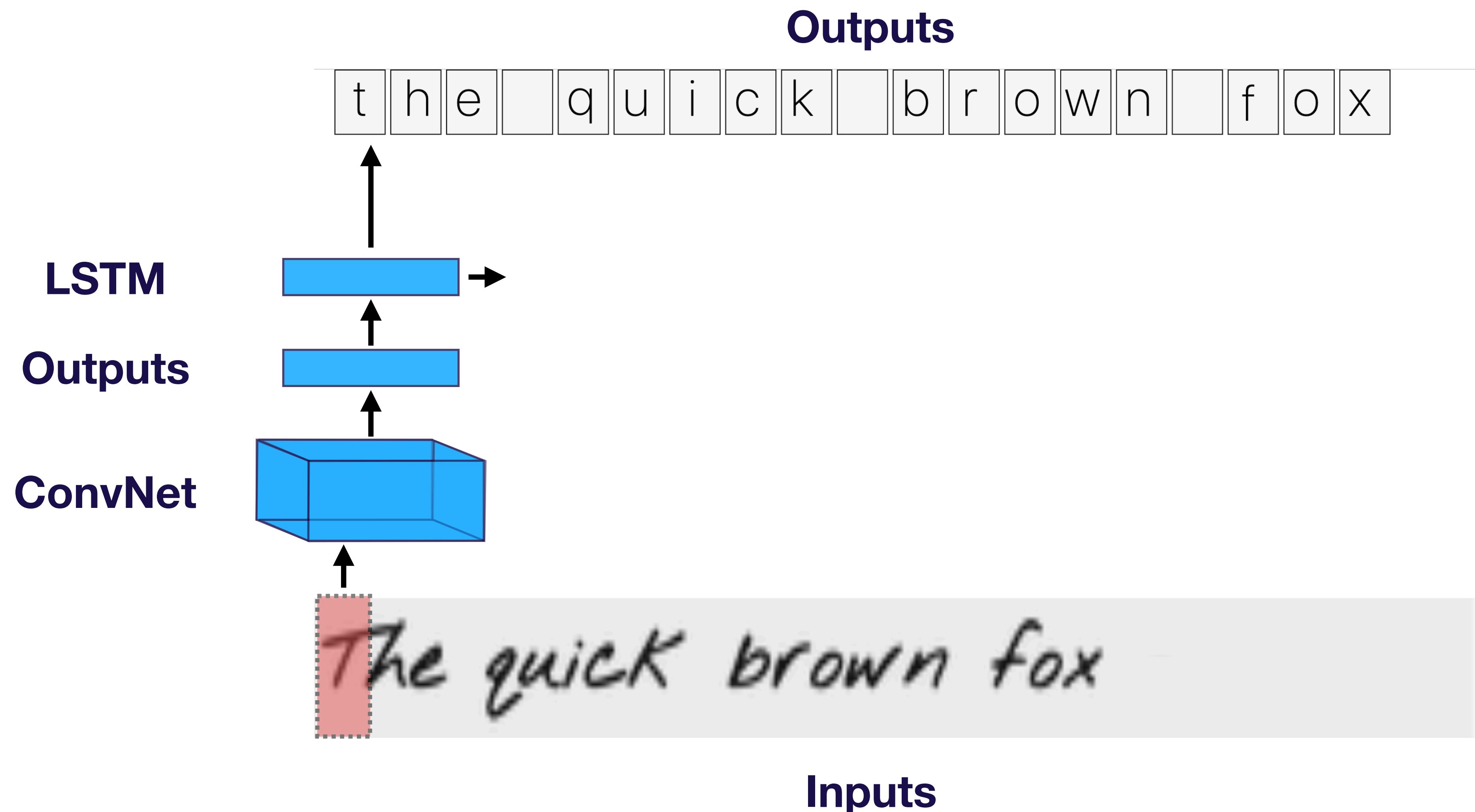
Challenge: different line spacing



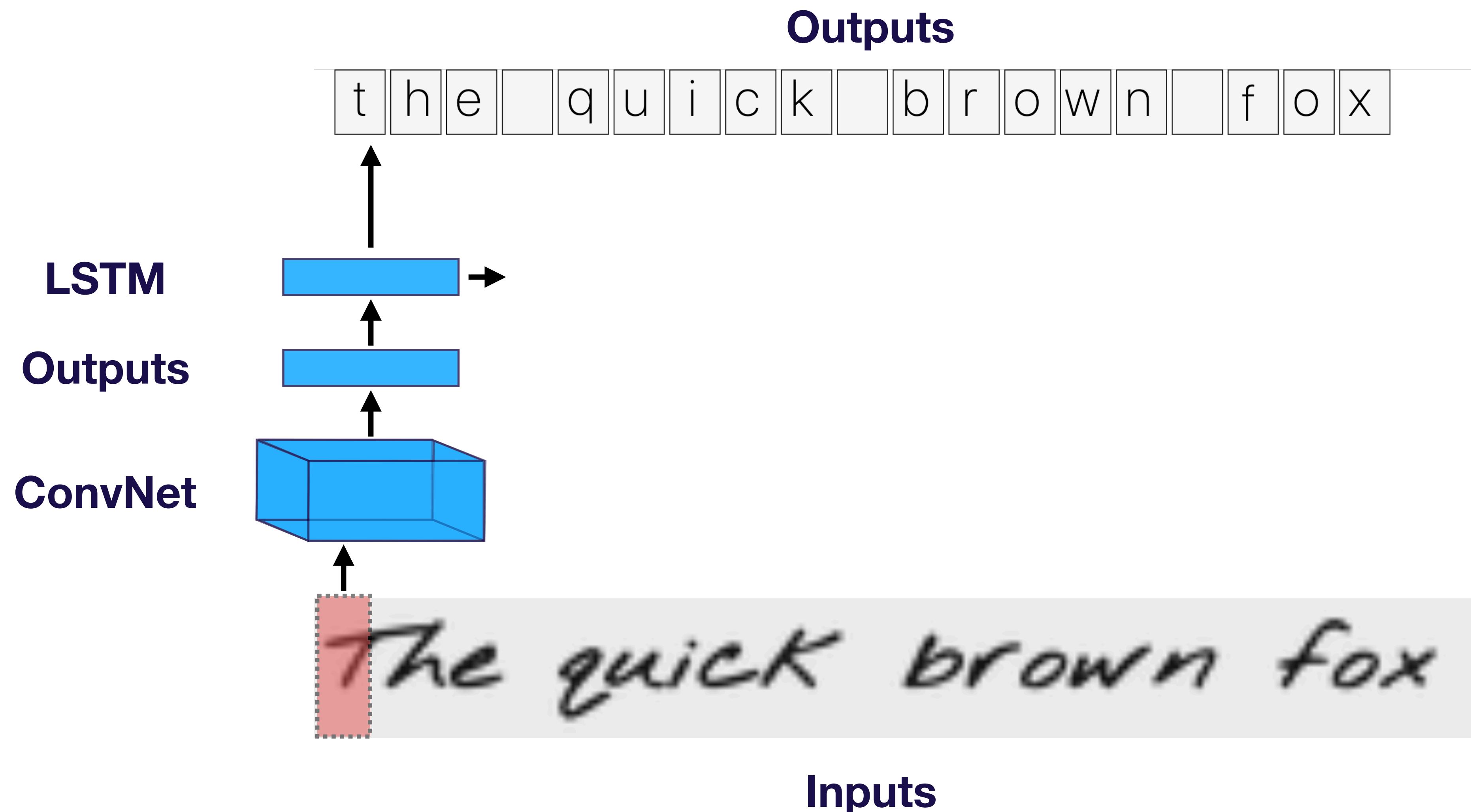
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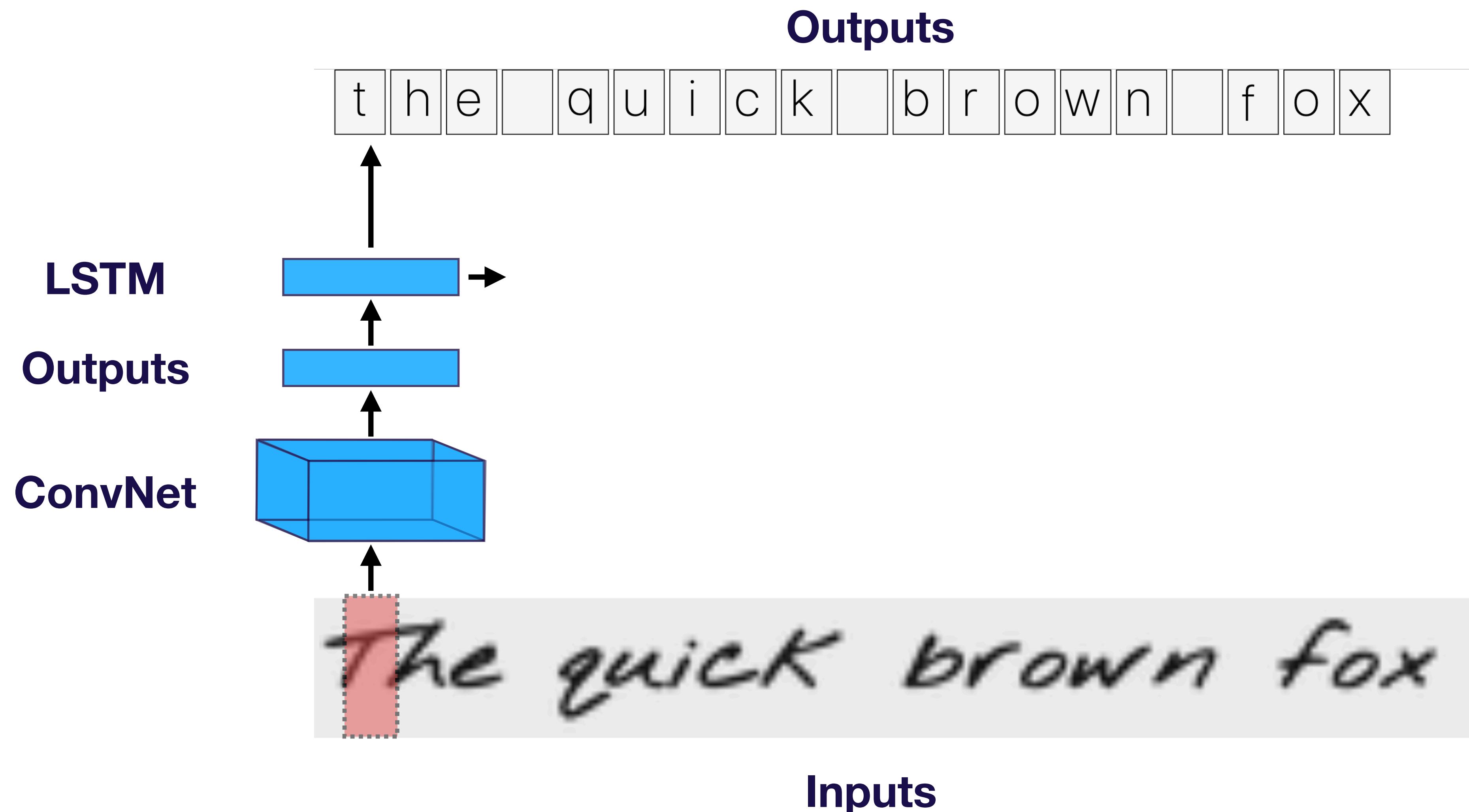
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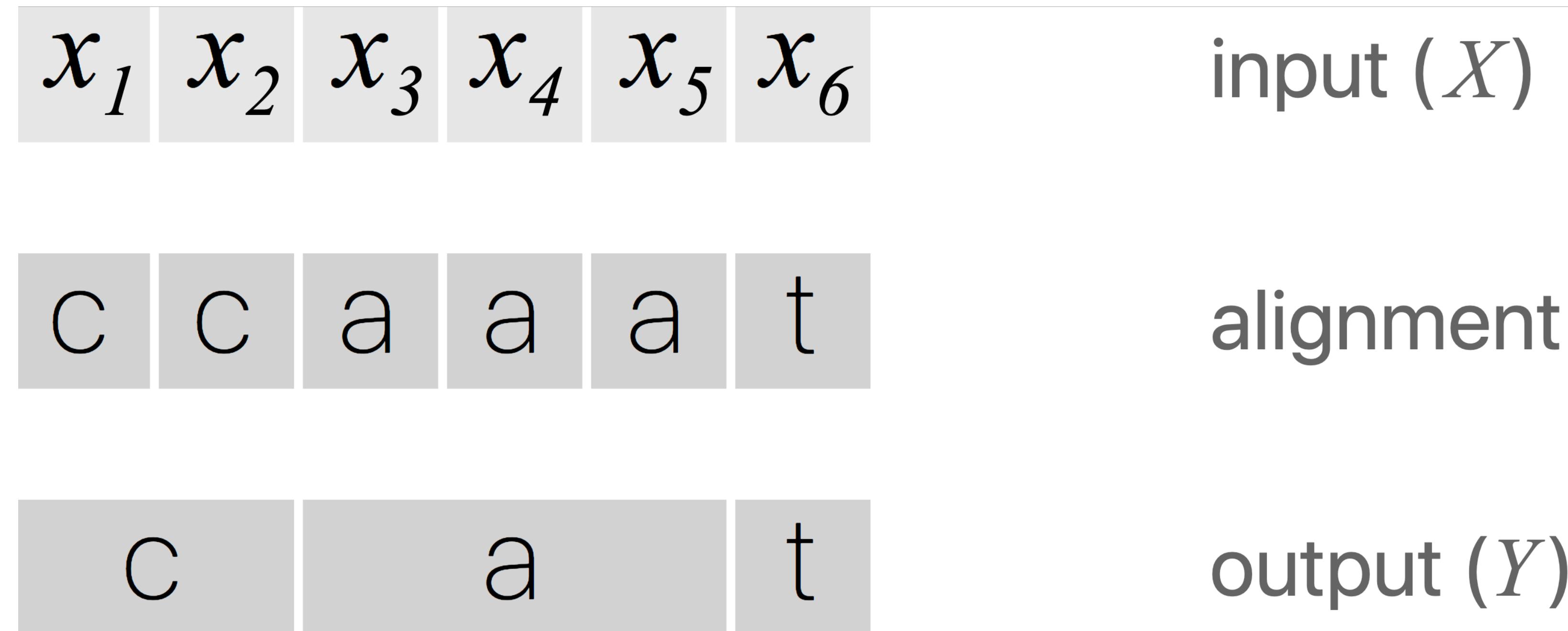
Challenge: different line spacing



Challenge: different line spacing



Intuition for the solution



Sequence Modeling with CTC, Awni Hannun, Distill 10.23915/distill.00008

CTC Loss

h h e ε ε | | | ε | | o

First, merge repeat
characters.

h e ε | ε | ε | o

Then, remove any ϵ
tokens.

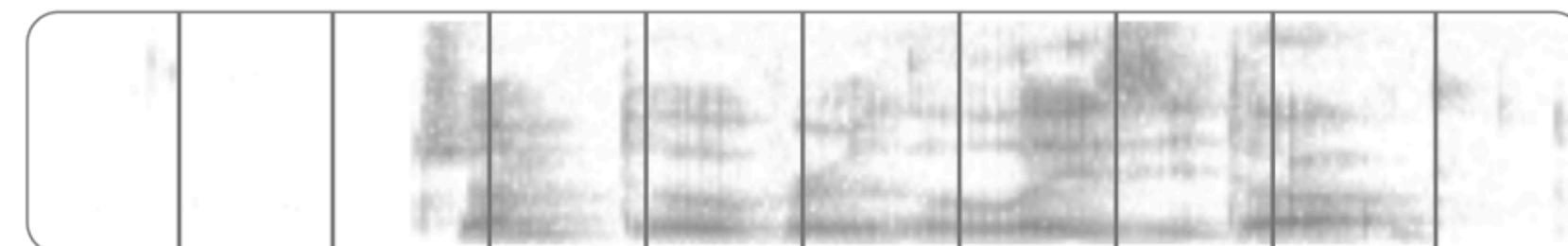
h e | | | o

The remaining characters
are the output.

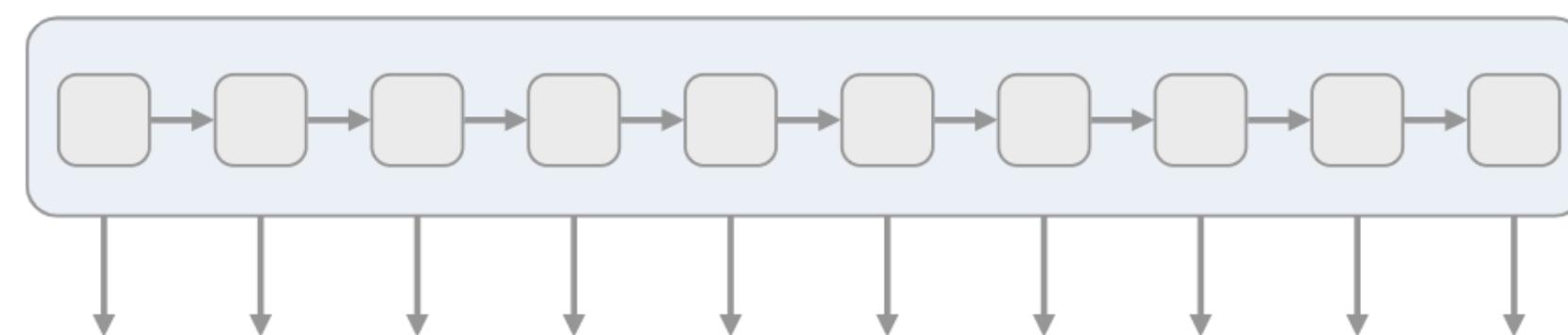
h e | | o

Sequence Modeling with CTC, Awni Hannun, Distill 10.23915/distill.00008

Training with the CTC loss



We start with an input sequence,
like a spectrogram of audio.



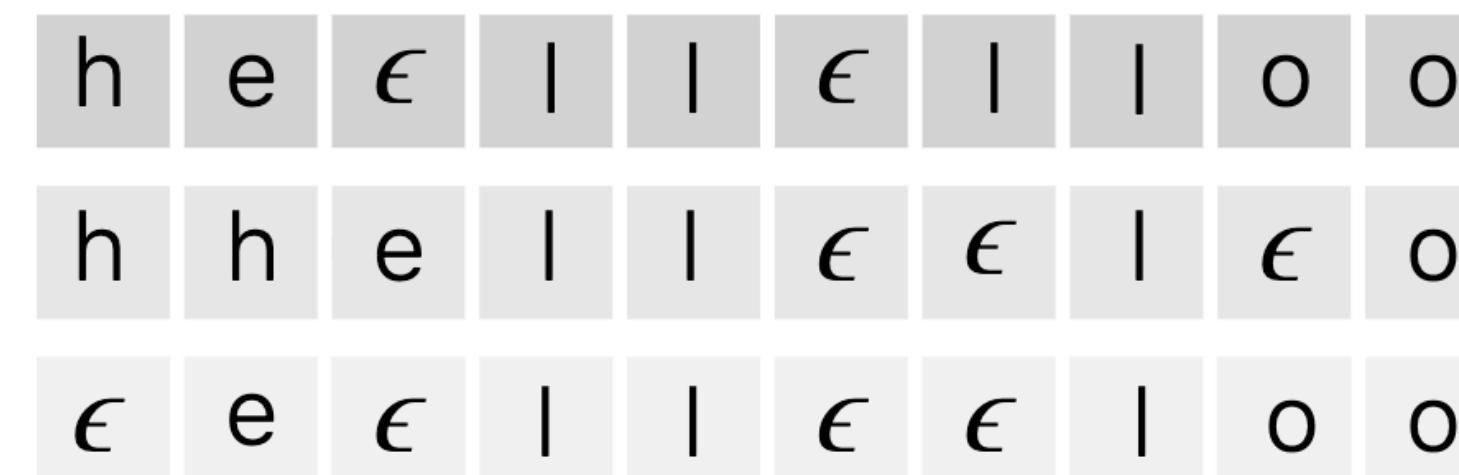
The input is fed into an RNN,
for example.

h	h	h	h	h	h	h	h	h	h
e	e	e	e	e	e	e	e	e	e
l	l	l	l	l	l	l	l	l	l
o	o	o	o	o	o	o	o	o	o
ε	ε	ε	ε	ε	ε	ε	ε	ε	ε

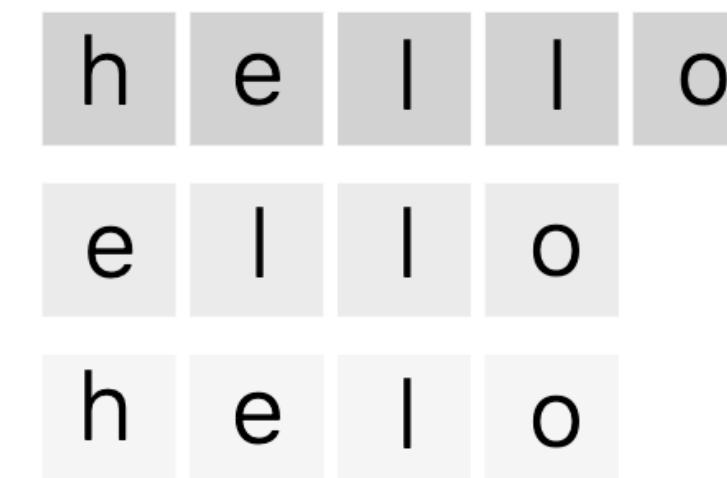
The network gives $p_t(a | X)$,
a distribution over the outputs
 $\{h, e, l, o, \epsilon\}$ for each input step.

Sequence Modeling with CTC, Awni Hannun, Distill 10.23915/distill.00008

Training with the CTC loss



With the per time-step output distribution, we compute the probability of different sequences



By marginalizing over alignments, we get a distribution over outputs

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For more of the technical details

Sequence Modeling with CTC, Awni Hannun, Distill 10.23915/distill.00008