

Machine Learning in PK/PD Modeling

Steven L. Shafer, MD

Professor of Anesthesiology, Perioperative and Pain Medicine, Stanford University
Adjunct Associate Professor of Bioengineering and Therapeutic Sciences, UCSF

Conflicts of Interest:

None



MIT TECHNOLOGY REVIEW

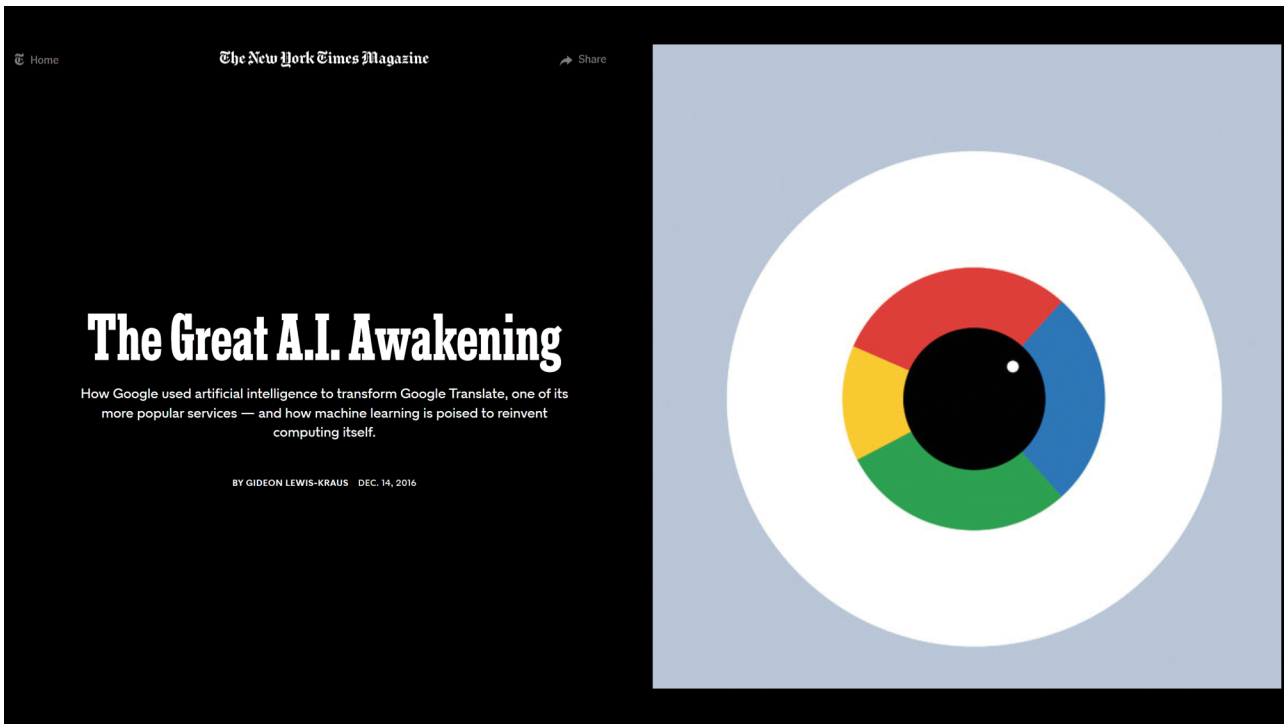
Business Impact

Google Stakes Its Future on a Piece of Software

Alphabet, number 5 on our list of the 50 Smartest Companies, thinks it can wrest the cloud computing market away from Amazon by helping companies make use of machine learning with a tool called TensorFlow.

by Tom Simonite June 27, 2017







Prof. Jun Rekimoto, Ph.D

Professor, Interfaculty Initiative in Information Studies,
The University of Tokyo

Kilimanjaro is a mountain of 19,710 feet covered with snow and is said to be the highest mountain in Africa. The summit of the west is called “Ngaje Ngai” in Masai, the house of God. Near the top of the west there is a dry and frozen dead body of leopard. No one has ever explained what leopard wanted at that altitude.

Kilimanjaro is a snow-covered mountain 19,710 feet high, and is said to be the highest mountain in Africa. Its western summit is called the Masai “Ngaje Ngai,” the House of God. Close to the western summit there is the dried and frozen carcass of a leopard. No one has explained what the leopard was seeking at that altitude.

This is an example of Google Translate. I have no idea if the translation is any good, but I thought I would test it and let you be the judge. Does the language seem natural?

Dit is een voorbeeld van Google Translate. Ik heb geen idee of de vertaling goed is, maar ik dacht dat ik het zou testen en je de rechter zou laten zijn. Lijkt de taal natuurlijk?

AlphaGo beats human Go champ in milestone for artificial intelligence



South Koreans watch a TV showing the historic match between Go champion Lee Sedol and AlphaGo at a Seoul train station. (Kim Hee-Chul / European Pressphoto Agency)

By **Steven Borowiec and Tracey Lien** - Contact Reporter

MARCH 12, 2016, 6:23 PM | REPORTING FROM SEOUL

First went checkers, then fell chess. Now, a computer program has defeated the world's top player in the ancient east Asian board game of Go — a major milestone for artificial intelligence that brings to a close the era of board games as benchmarks in computing.

Monte Carlo tree search of neural networks.

Neural networks bootstrapped from human gameplay expertise.

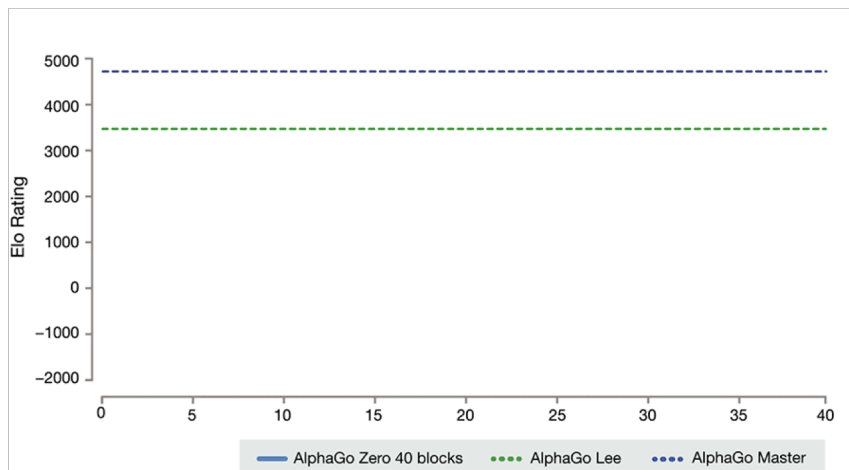
Trained further by playing games against other instances of itself



nature.com

354 | NATURE | VOL 550 | 19 October 2017

Mastering the game of Go without human knowledge



Google's AlphaZero Destroys Stockfish In 100-Game Match

Mike Klein

Dec 6, 2017, 12:50 PM

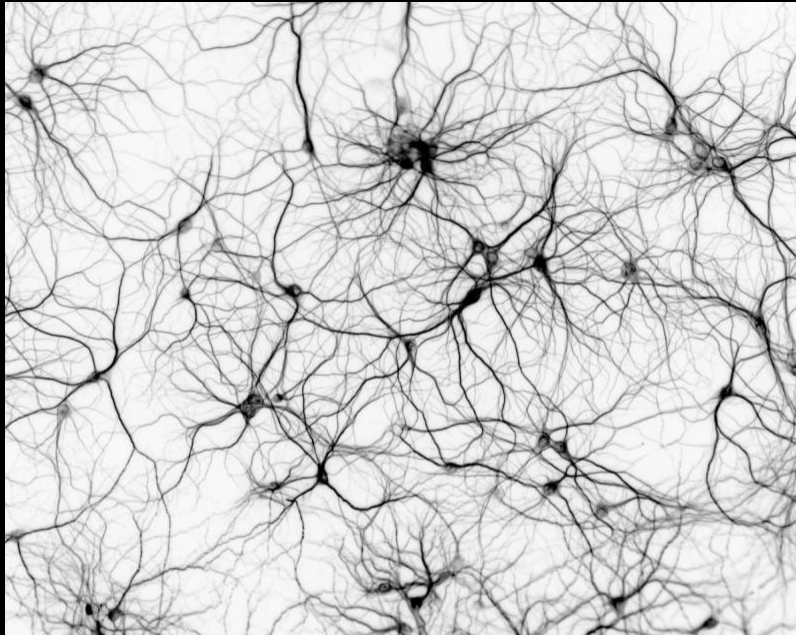
Chess changed forever today. And maybe the rest of the world did, too.

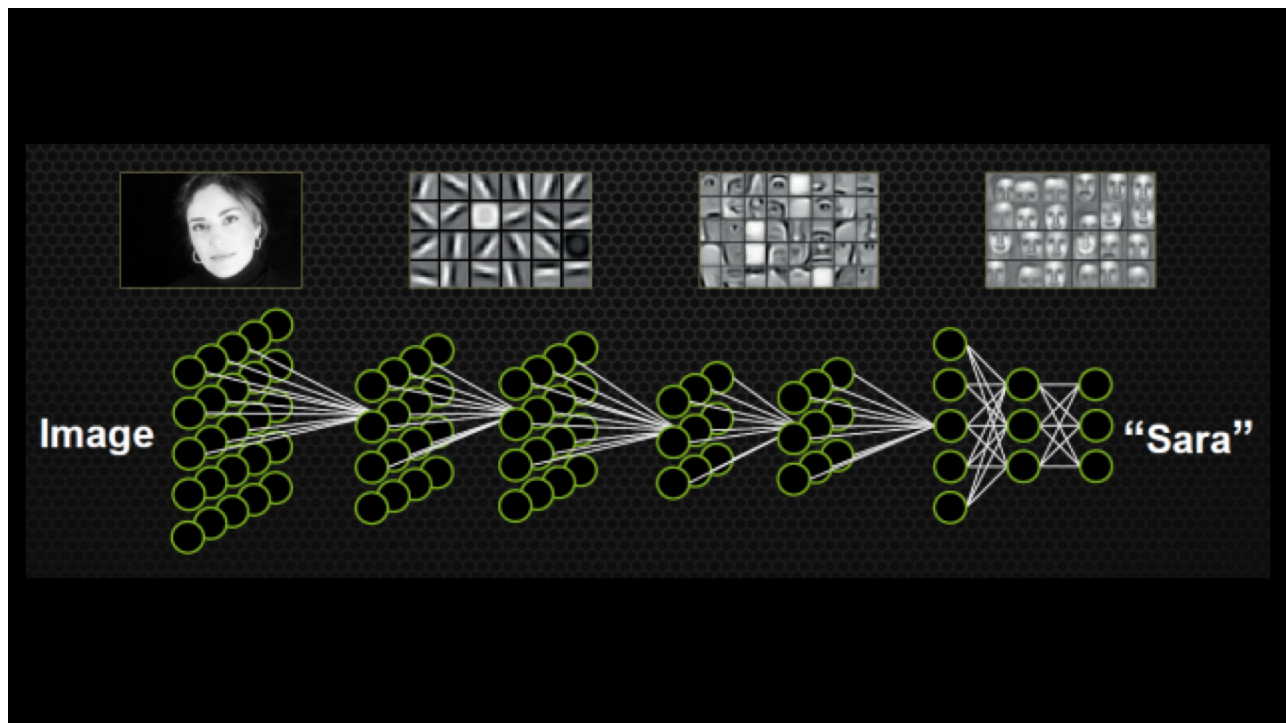
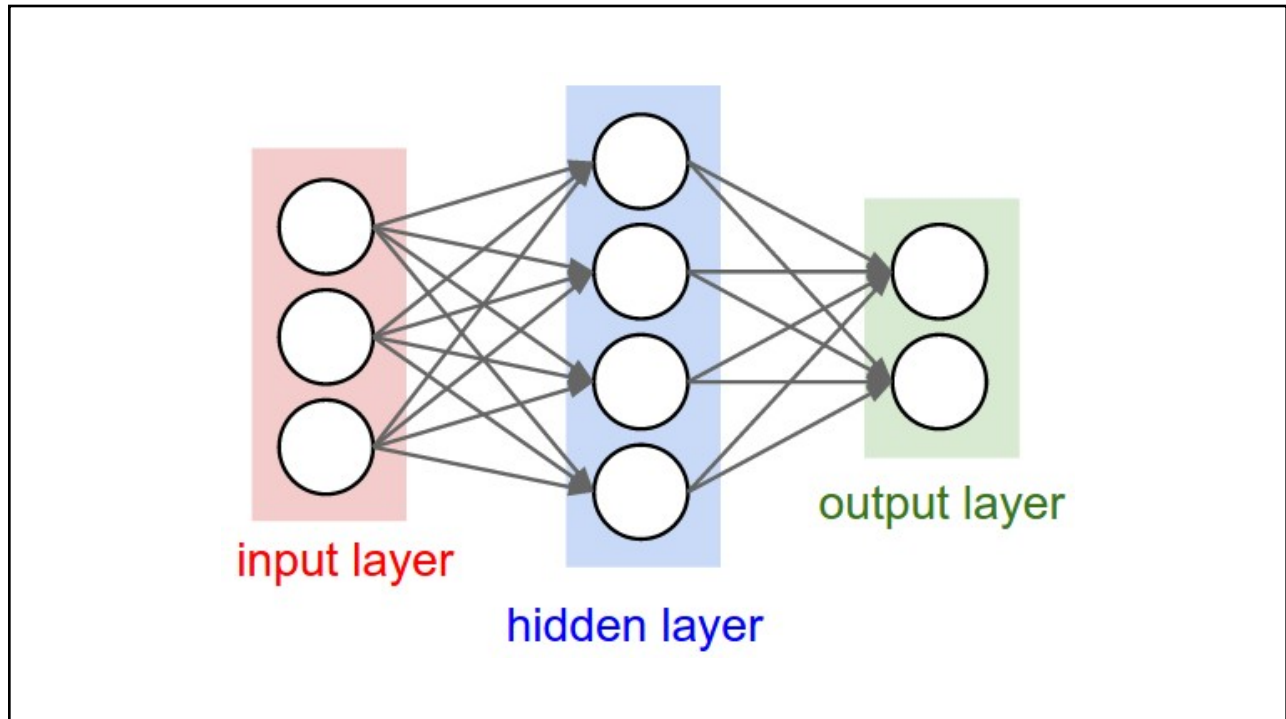
A little more than a year after AlphaGo sensationally won against the top Go player, the artificial-intelligence program AlphaZero has obliterated the highest-rated chess engine.

Stockfish, which for most top players is their go-to preparation tool, and which won the 2016 TCEC Championship and the 2017 Chess.com Computer Chess Championship, didn't stand a chance. AlphaZero won the closed-door, 100-game match with 28 wins, 72 draws, and zero losses.

Oh, and it took AlphaZero only four hours to "learn" chess. Sorry humans, you had a good run.

That's right -- the programmers of AlphaZero, housed within the DeepMind division of Google, had it use a type of "machine learning," specifically reinforcement learning. Put more plainly, AlphaZero was not "taught" the game in the traditional sense. That means no opening book, no endgame tables, and apparently no complicated algorithms dissecting minute differences between center pawns and side pawns.





GPUs Power World's Largest Deep Learning Algorithm

2012
Google Datacenter



1000 CPU Servers

1.7 billion parameter
neural network

Today
NVIDIA & Stanford AI Lab

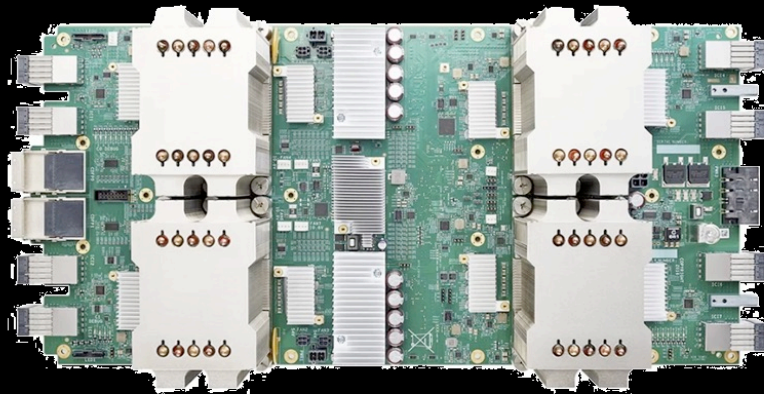


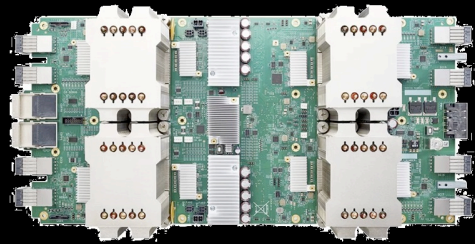
16 GPU-Accelerated Servers

11.2 billion parameter
neural network

6.5x Bigger Artificial Neural Network

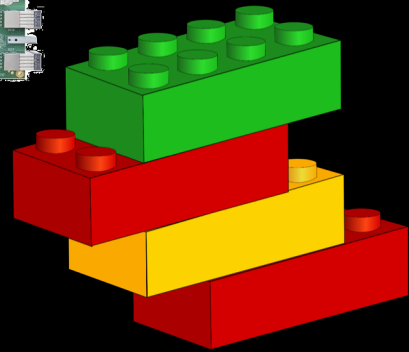
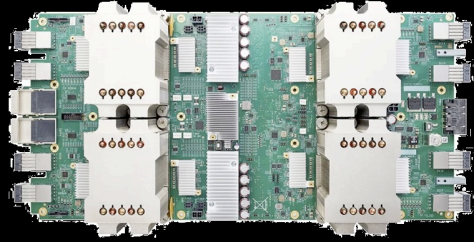
Tensor Processing Unit





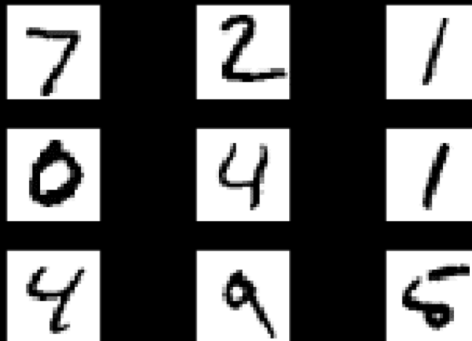


TensorFlow™

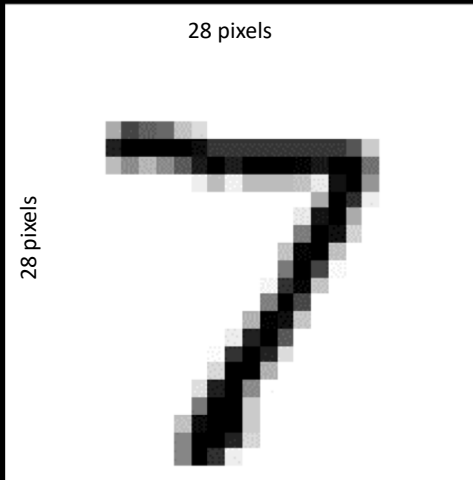


MNIST EXAMPLE

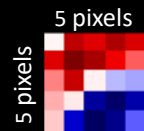
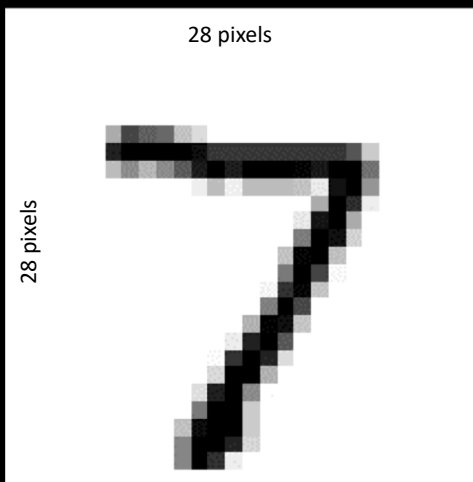
- National Institute of Standards and Technology
- 60,000 handwritten digits, 0-9



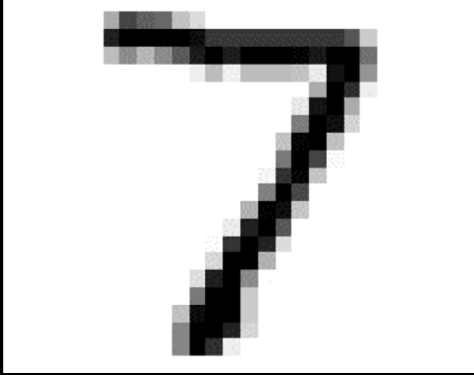
Convolution Neural Network



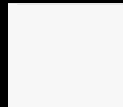
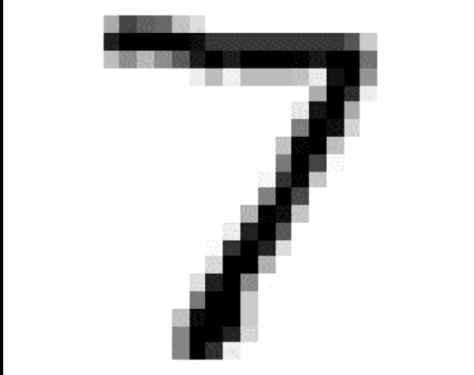
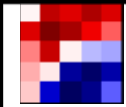
Convolution Neural Network



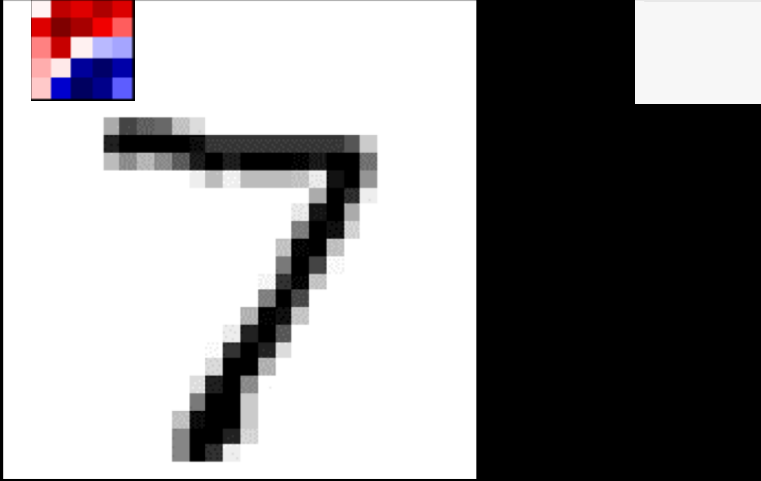
Convolution Neural Network



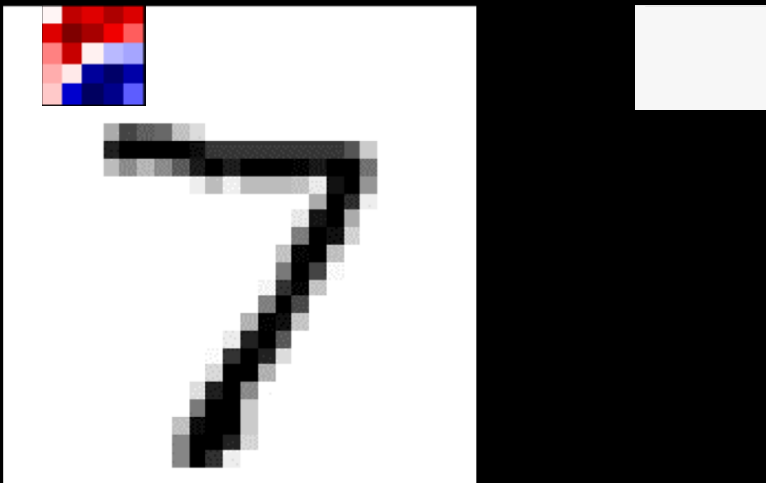
Convolution Neural Network



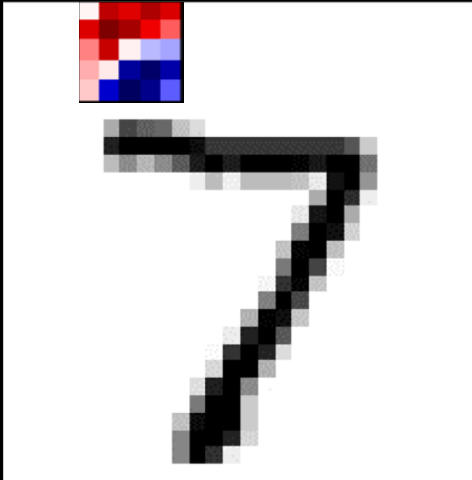
Convolution Neural Network



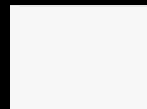
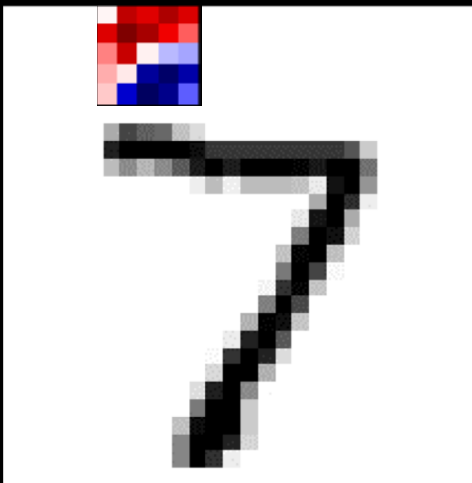
Convolution Neural Network



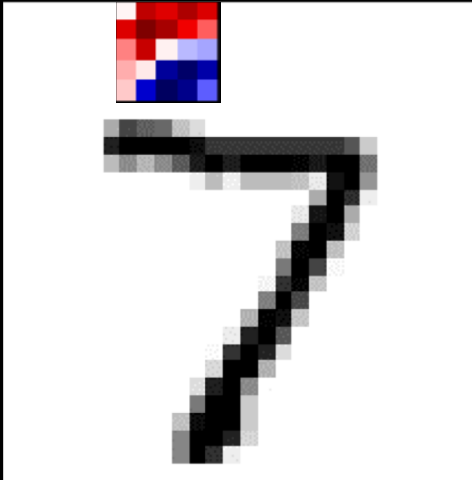
Convolution Neural Network



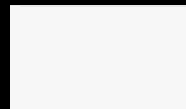
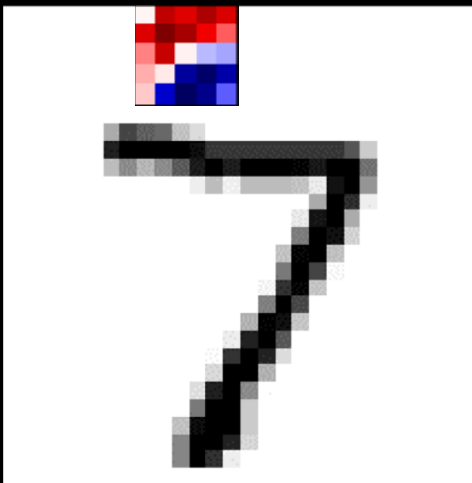
Convolution Neural Network



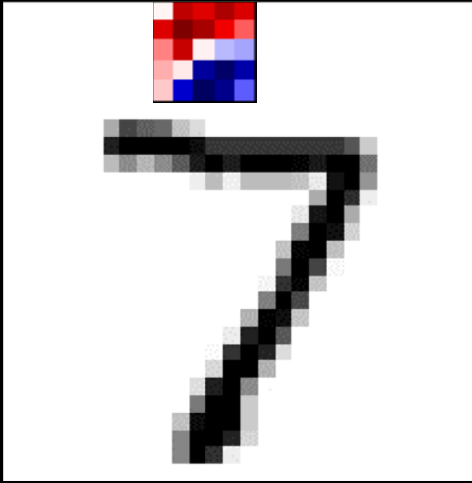
Convolution Neural Network



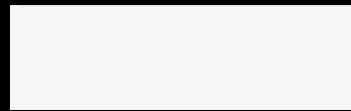
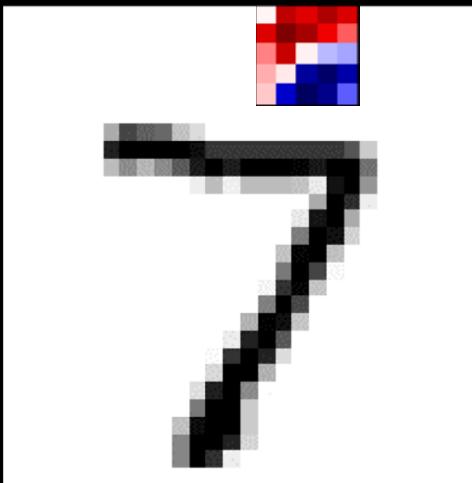
Convolution Neural Network



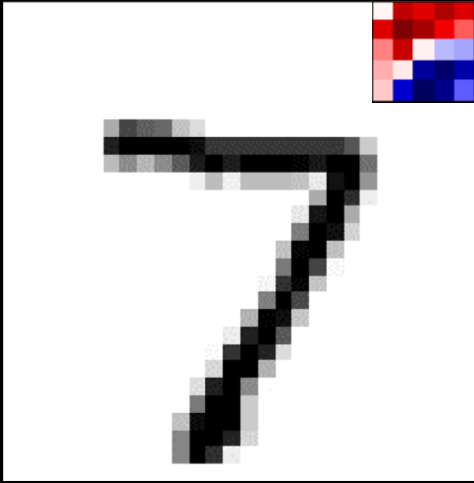
Convolution Neural Network



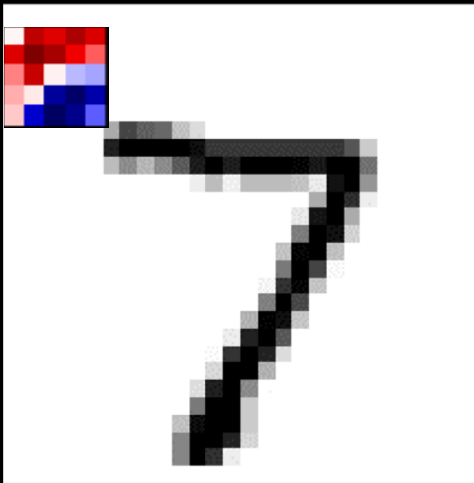
Convolution Neural Network



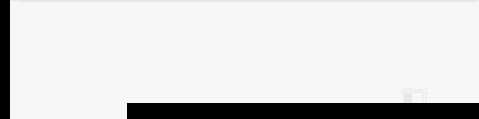
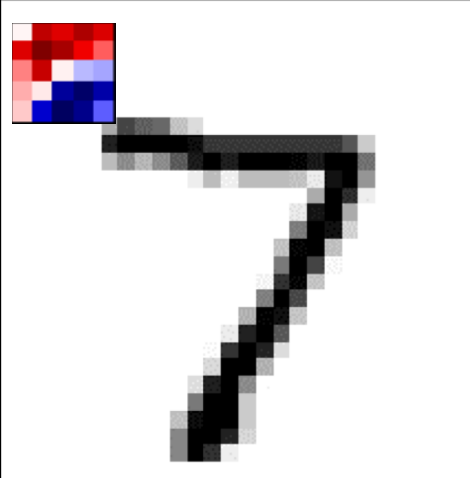
Convolution Neural Network



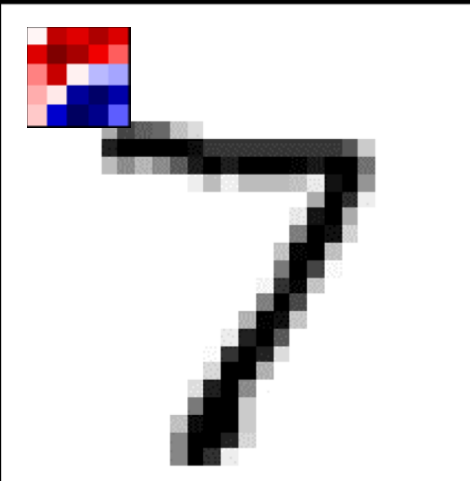
Convolution Neural Network



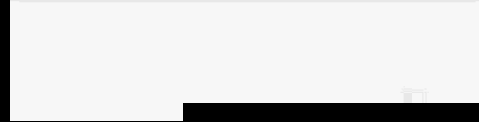
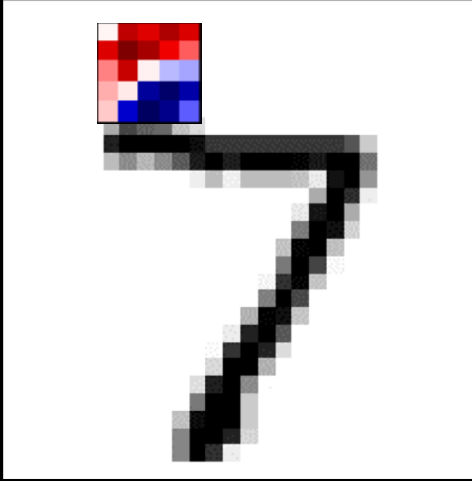
Convolution Neural Network



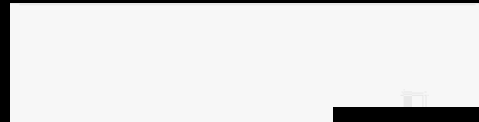
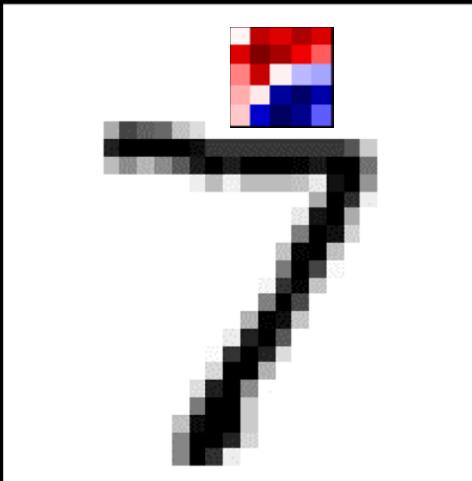
Convolution Neural Network



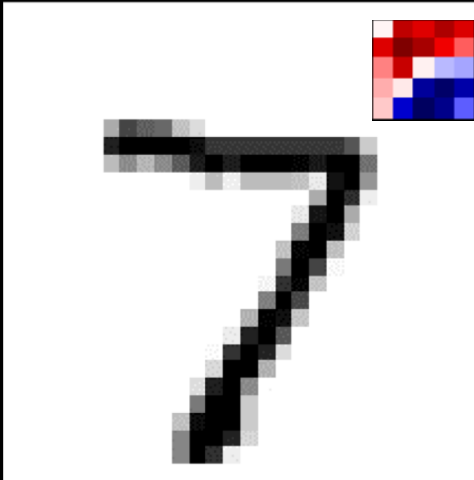
Convolution Neural Network



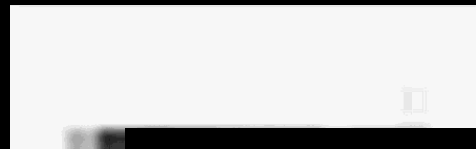
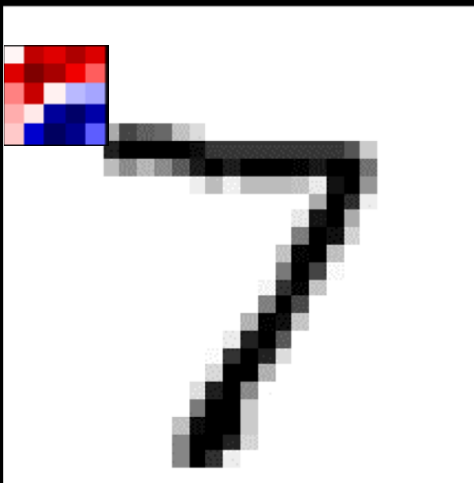
Convolution Neural Network



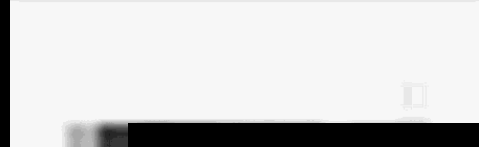
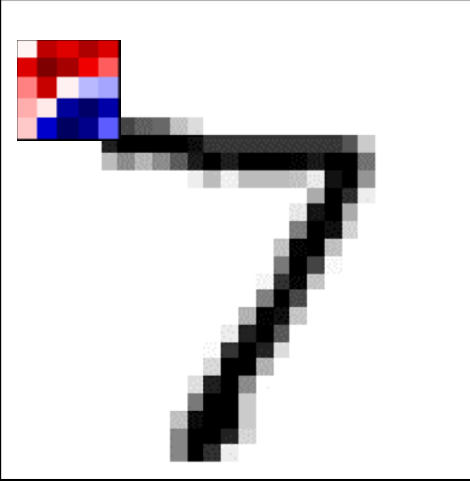
Convolution Neural Network



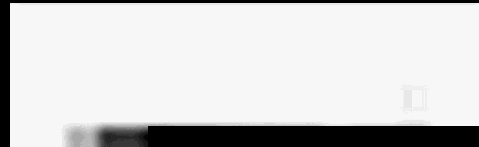
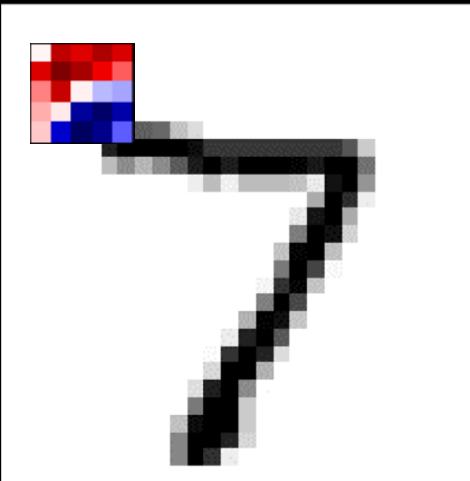
Convolution Neural Network



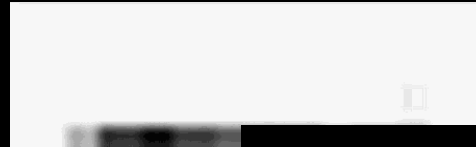
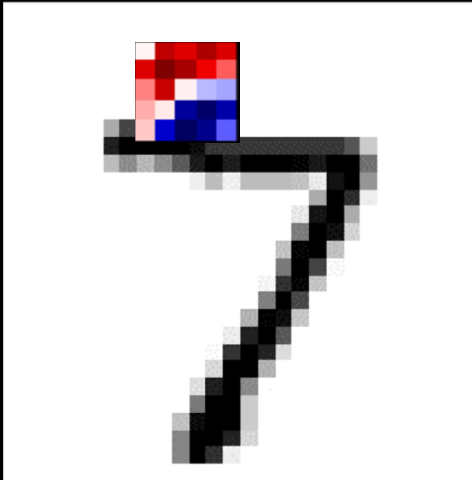
Convolution Neural Network



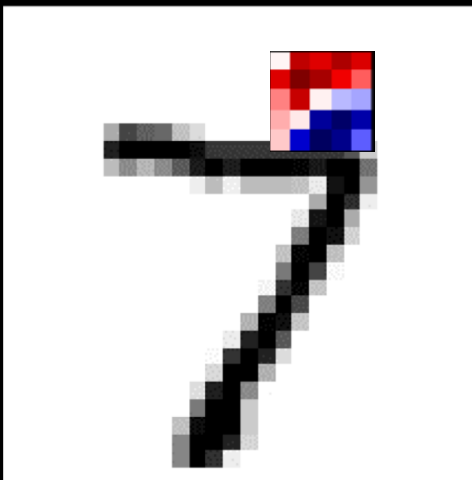
Convolution Neural Network



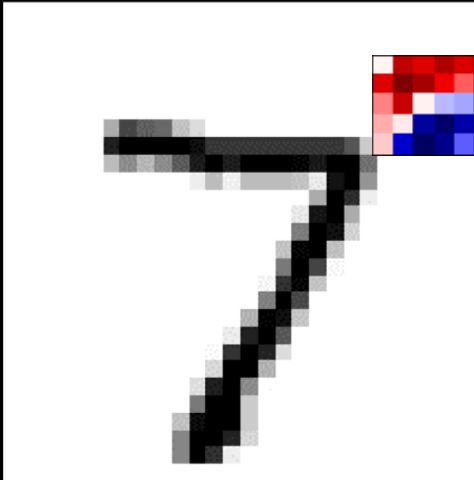
Convolution Neural Network



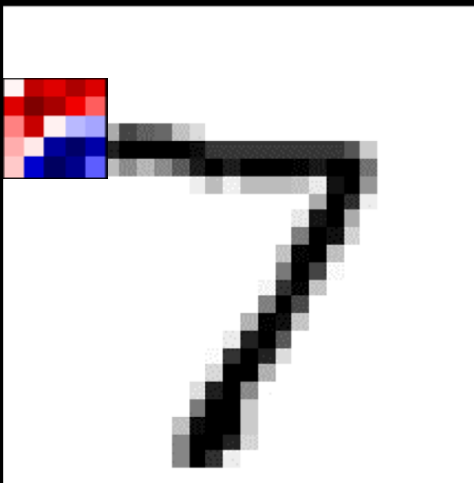
Convolution Neural Network



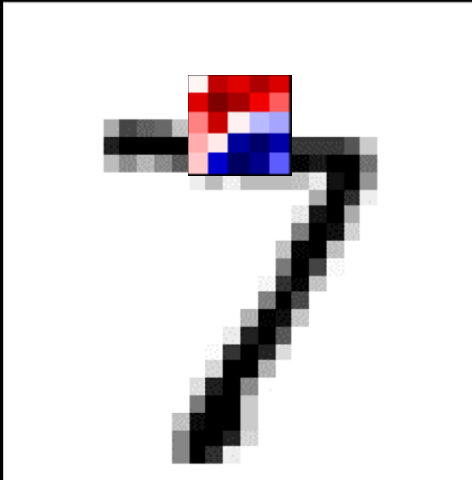
Convolution Neural Network



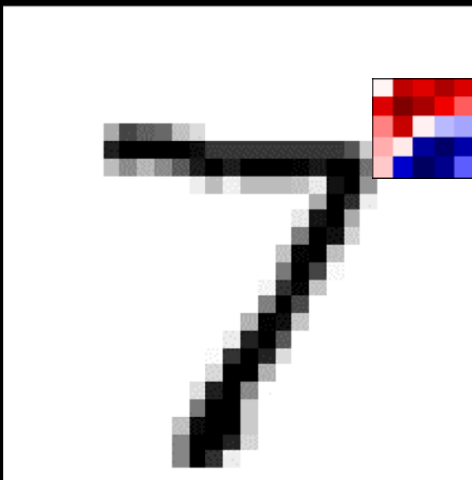
Convolution Neural Network



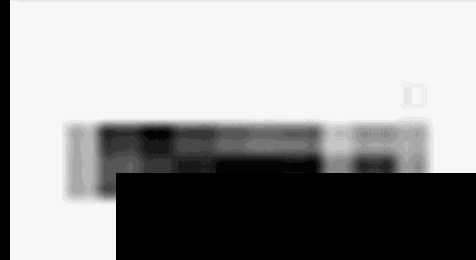
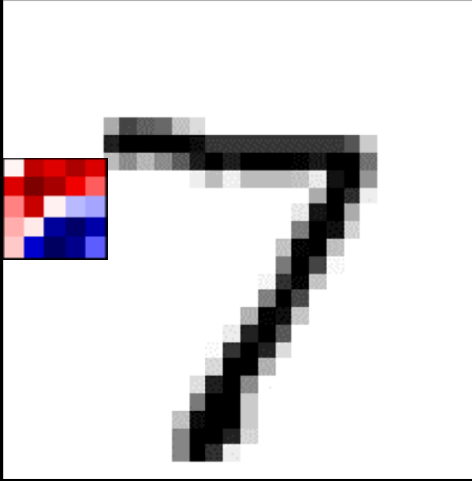
Convolution Neural Network



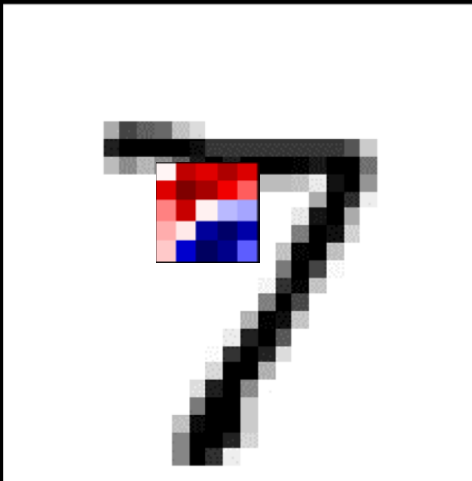
Convolution Neural Network



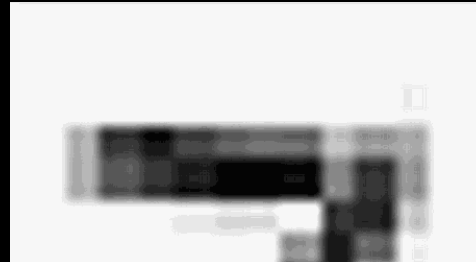
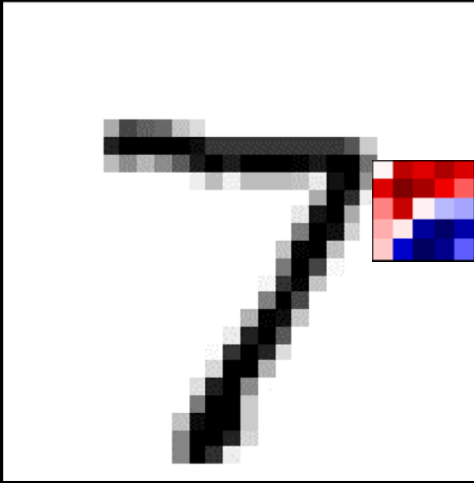
Convolution Neural Network



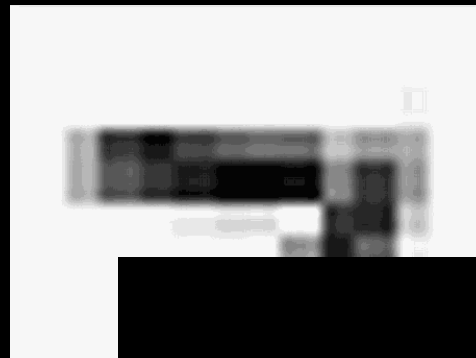
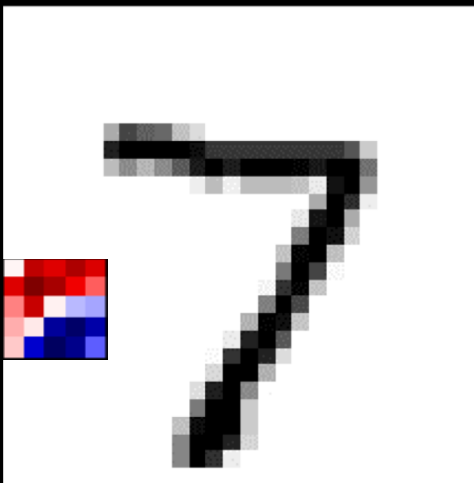
Convolution Neural Network



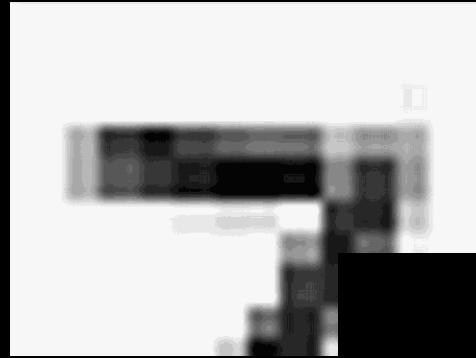
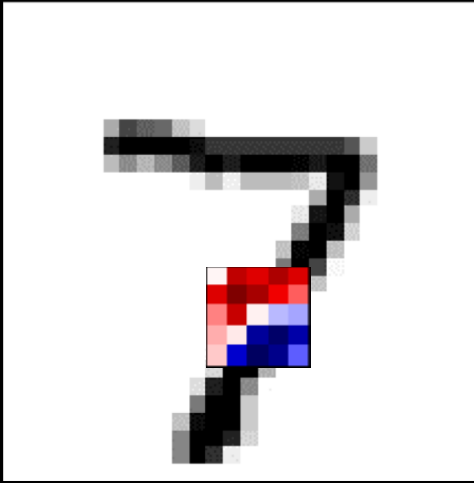
Convolution Neural Network



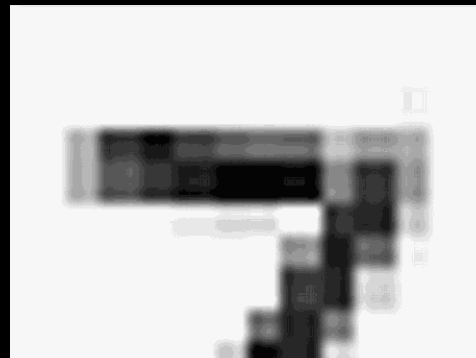
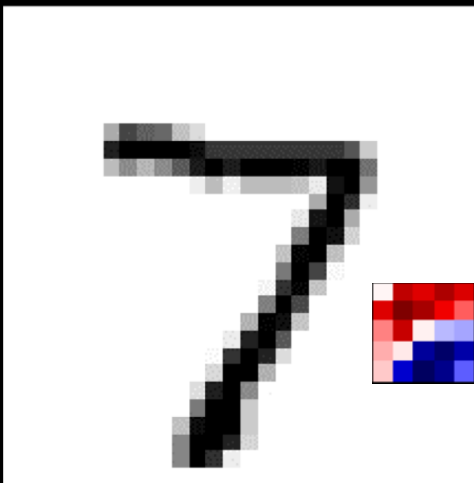
Convolution Neural Network



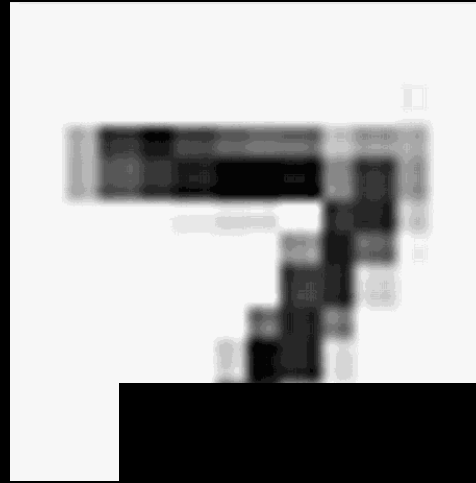
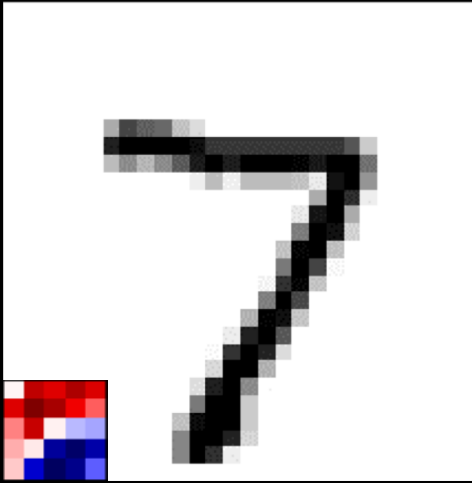
Convolution Neural Network



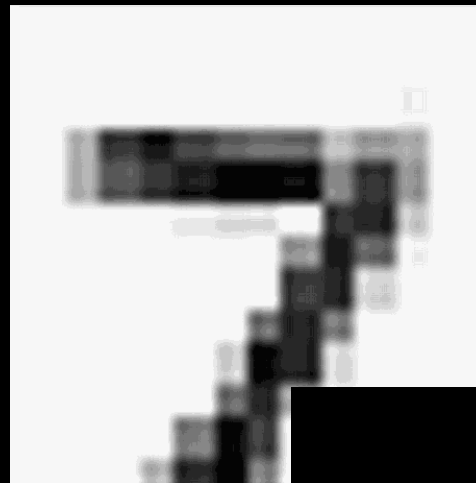
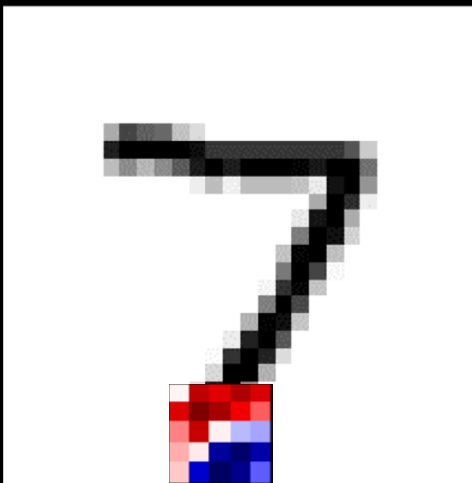
Convolution Neural Network



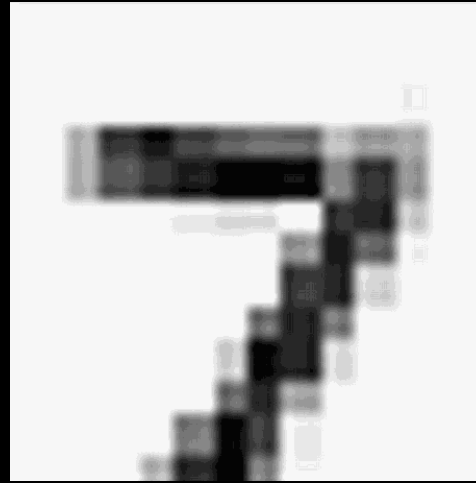
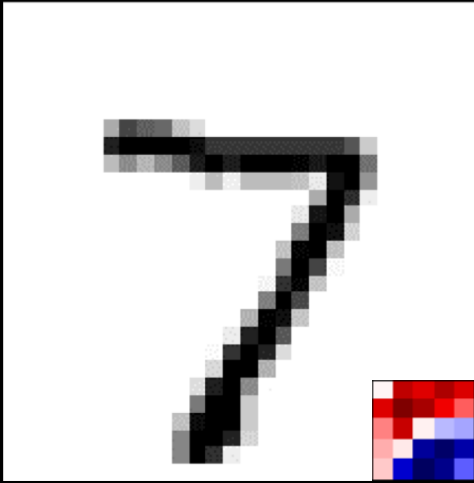
Convolution Neural Network



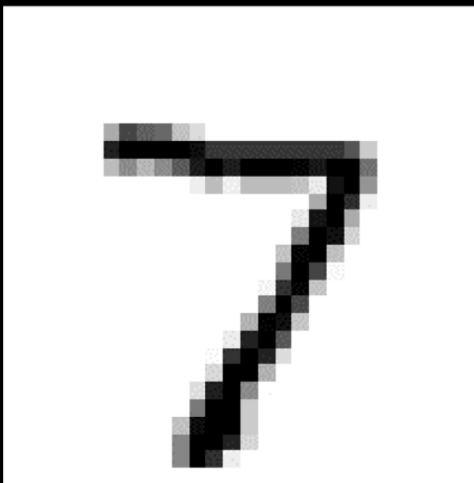
Convolution Neural Network



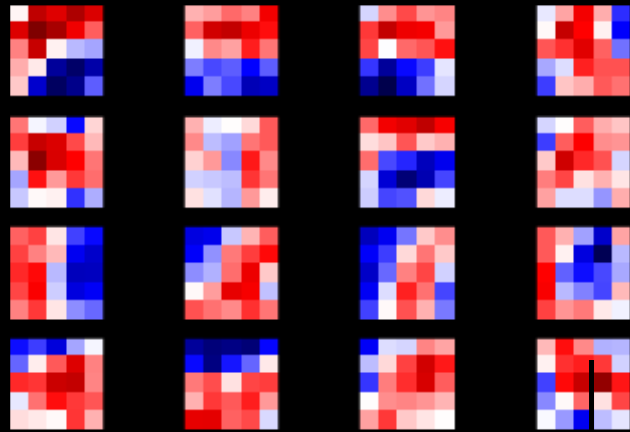
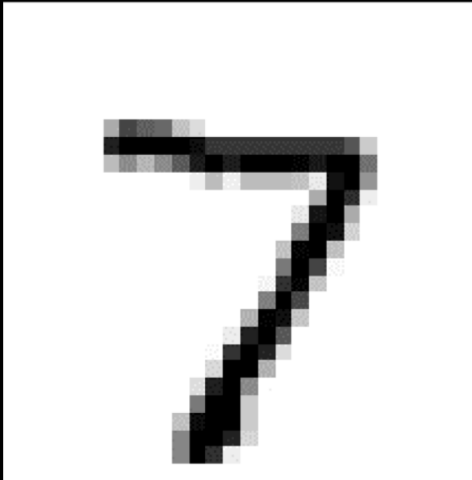
Convolution Neural Network



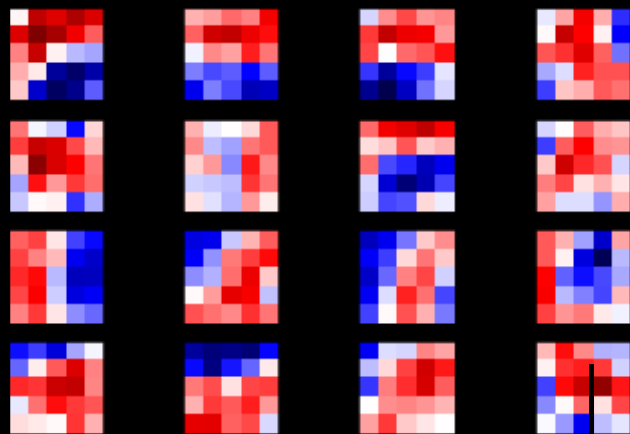
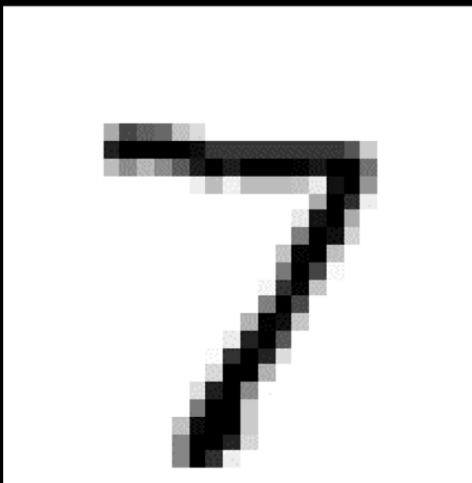
Convolution Neural Network



First Convolution Layer: 1 input \rightarrow 16 outputs

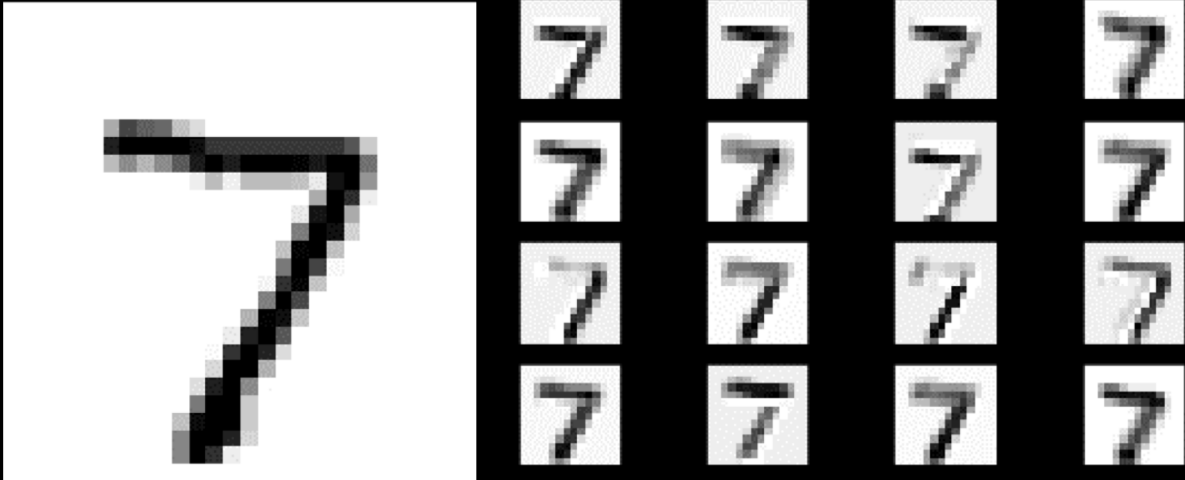


First Convolution Layer: 1 input \rightarrow 16 outputs

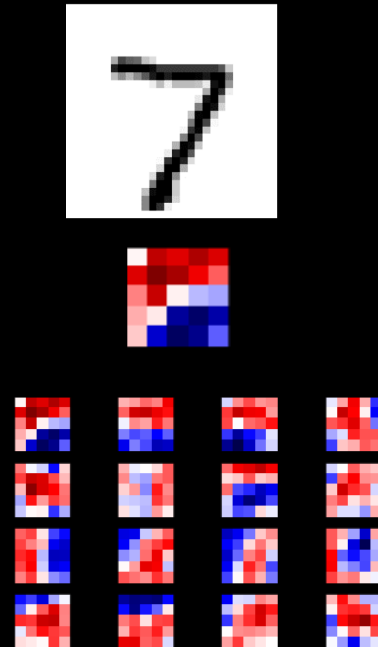


16 kernels x 25 weights/kernel = **400** weights

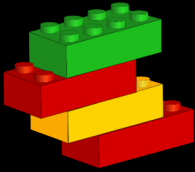
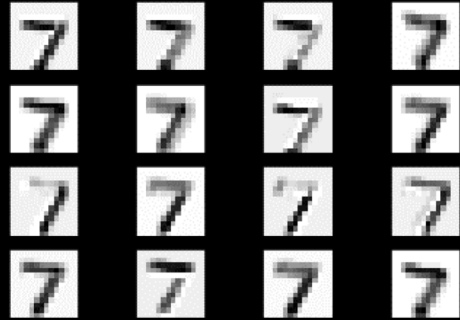
First Convolution Layer: 1 input \rightarrow 16 outputs



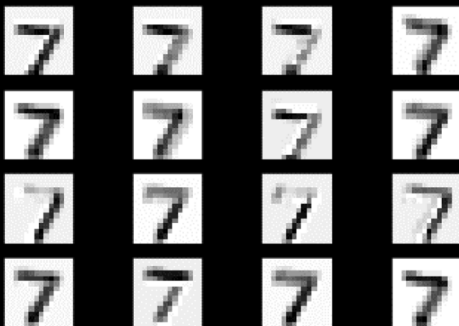
```
# First convolution layer
first_layer = new_conv_layer(
    input=image,
    num_input_channels=1,
    kernel_size=5,
    num_kernels=16)
```




```
# First convolution layer
first_layer = new_conv_layer(
    input=image,
    num_input_channels=1,
    kernel_size=5,
    num_kernels=16)
```



Second Convolution Layer

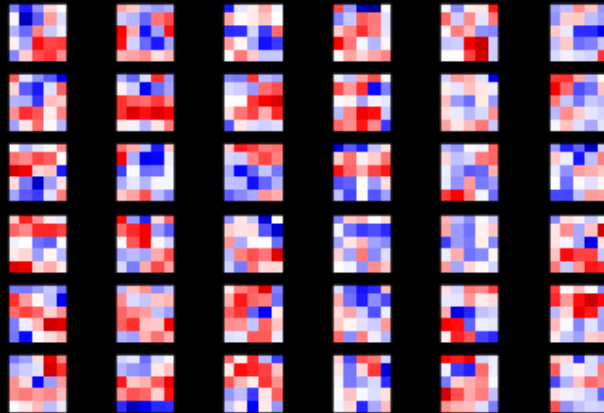


5x5 kernel

16 14x14 images

Second Convolution Layer

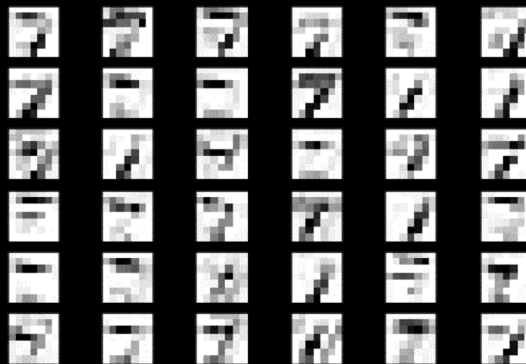
7



36 5x5 kernels = 900 weights

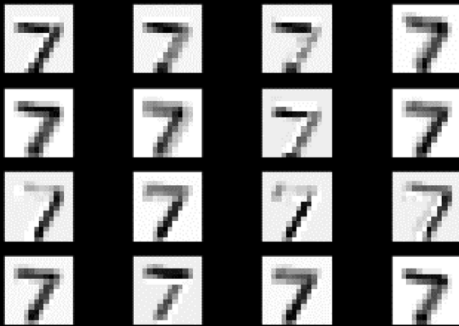
Second Convolution Layer

7

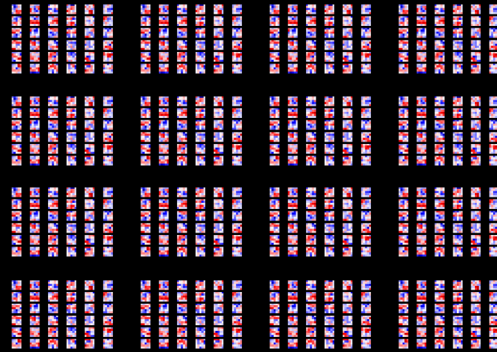


Each kernel produces an image

Second Convolution Layer



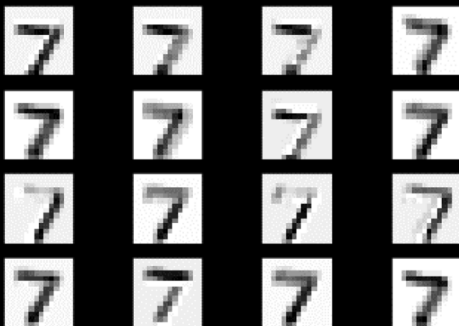
16 images in the input layer



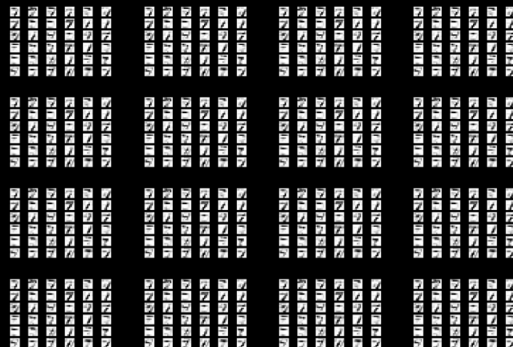
900 weights/image x 16 images = **14400** weights

Second Convolution Layer

16 inputs → 576 outputs


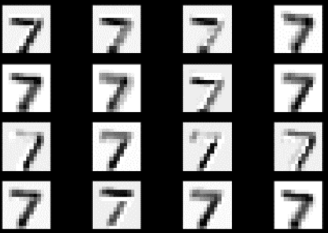

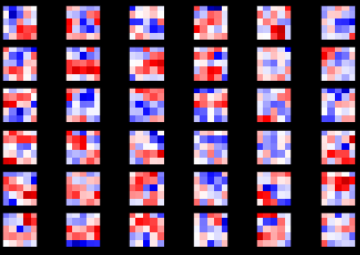


16 14x14 images



16 images x 36 kernels/image = 576 7x7 images

Second convolution layer
 second_layer = new_conv_layer(
 input=first_layer,
 num_input_channels=16,
 kernel_size=5,
 num_kernels=36)

Third Step – Flatten

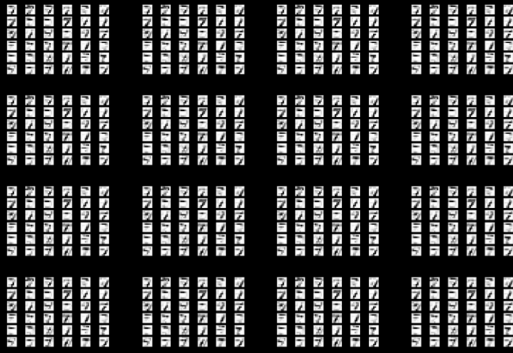



576 7x7 images 36 7x7 images = 1764 “features”

0 new weights

Third Step – Flatten

576 inputs \rightarrow 36 outputs



576 7x7 images

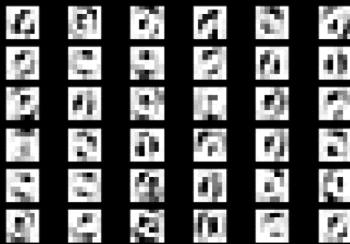


36 7x7 images = 1764 "features"

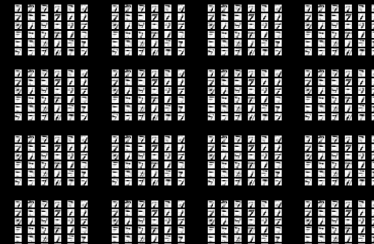
0 new weights

Third step – flatten

third_layer = **flatten_layer(second_layer)**

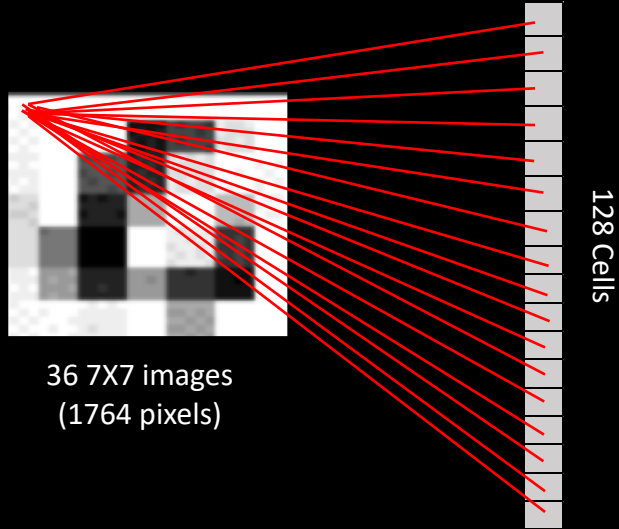


36 7x7 images = 1764 "features"

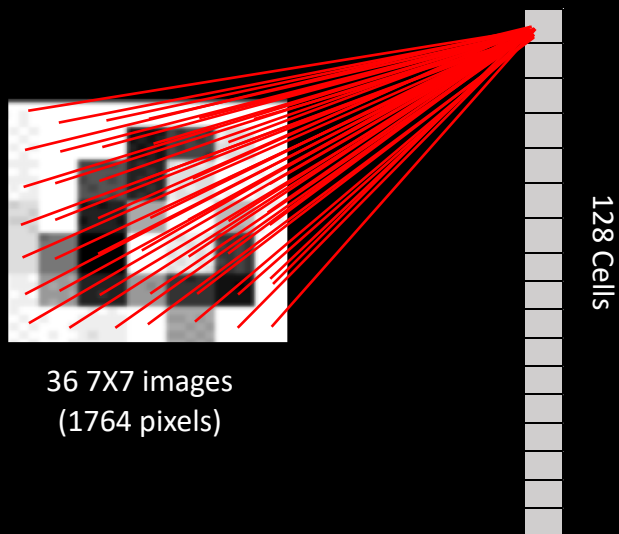


576 7x7 images

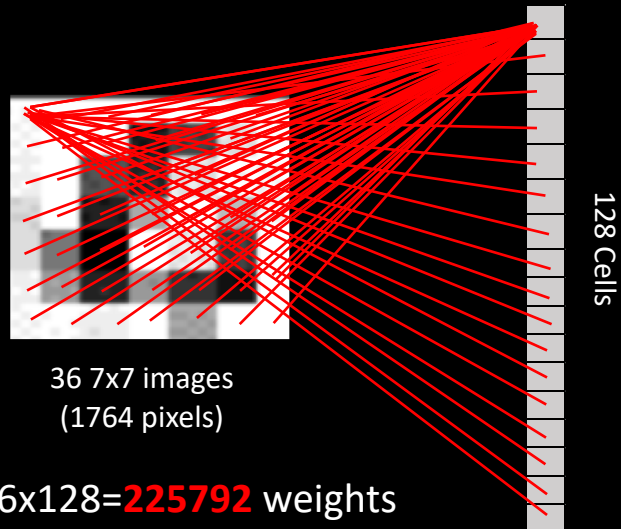
Fourth step: Fully Connected Layer



Fourth step: Fully Connected Layer

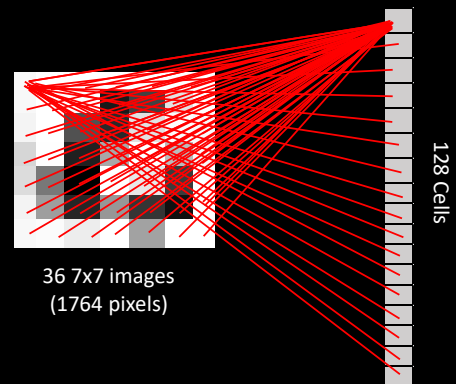
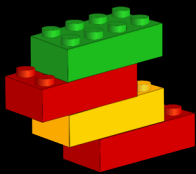


Fourth step: Fully Connected Layer

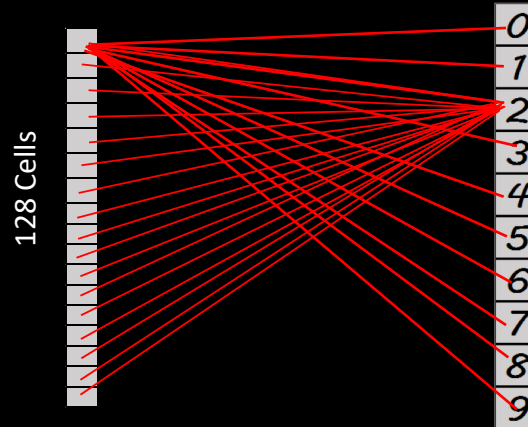


Fourth layer - the fully connected layer

```
fourth_layer = new_fc_layer(
    input=third_layer,
    num_inputs=1764,
    num_outputs=128)
```



Final Fully Connected Layer
128 inputs → 10 outputs

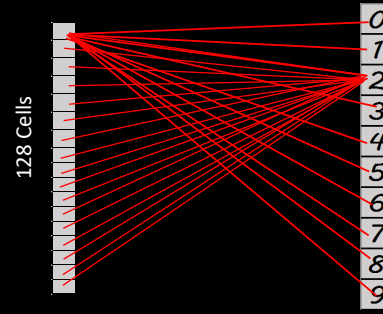


1280 weights

```
# Fifth layer – final fully connected layer
fifth_layer = new_fc_layer(
    input=fourth_layer,
    num_inputs=128
    num_outputs=10)
```


Fifth layer – final fully connected layer

```
fifth_layer = new_fc_layer(
    input=fourth_layer,
    num_inputs=128,
    num_outputs=10)
```



Lego Blocks of a Convolution Neural Network

Overall structure

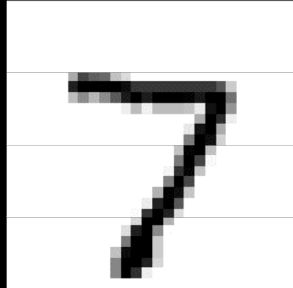
- 1 input → 10 outputs
- 241,872 weights

The steps were shown for image 1.

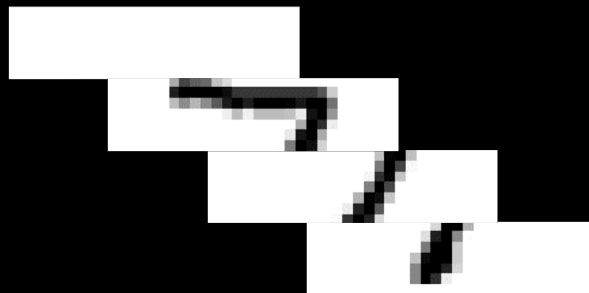
The steps run in parallel for all 60,000 images.

The 241,872 weights in the model are optimized for recognition of the entire data set.

Think of the “Image” as a time series



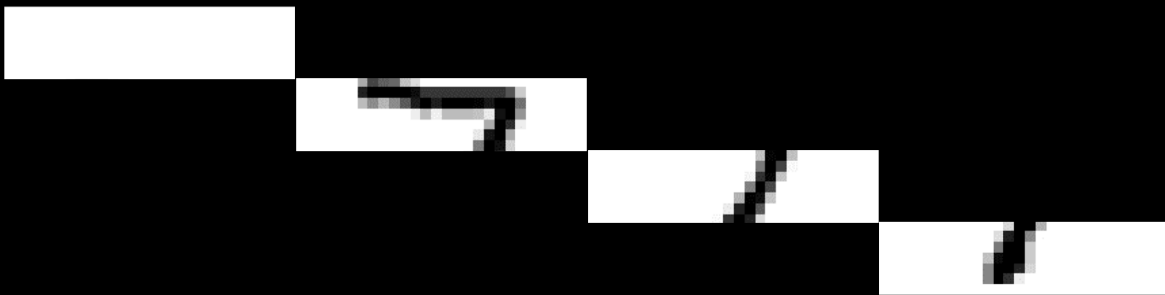
Think of the “Image” as a time series



Think of the “Image” as a time series



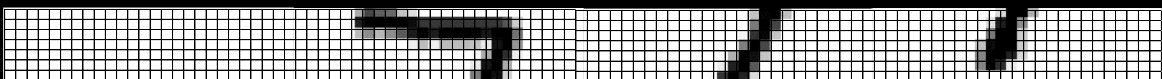
Think of the “Image” as a time series



Think of the “Image” as a time series



Our “image” has become a time series with 7 data channels (drug levels, vital signs, lab tests, etc)

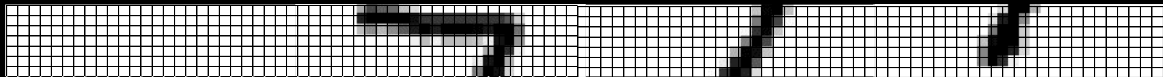


112 time units



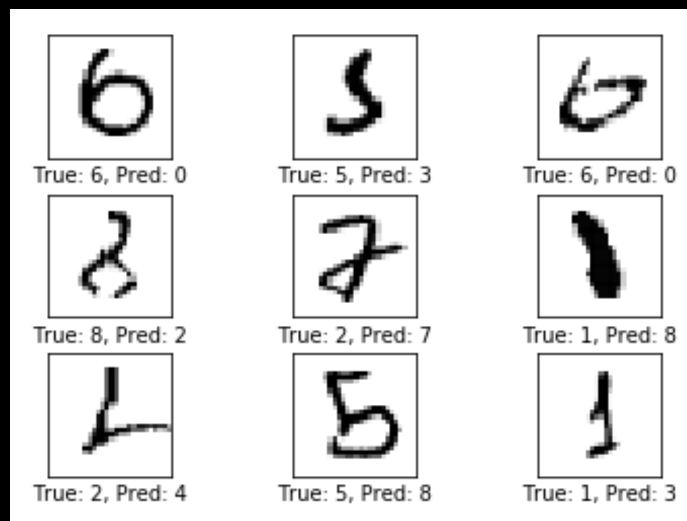
And 112 time units (months, clinic visits, etc)

7 data channels



112 time units

30,000 Iterations required 4 minutes
Predicted digits with 99% accuracy



PK/PD Modeling with CNN

- Input

- Linear array of cells, 1 per unit time
 - E.g. 1440 cells = 1 day
 - Each cell has the dose given at that minute

- Output

- Observed concentration at each minute (e.g., 1440 cells=1 day)
- 0 at each minute with no observation

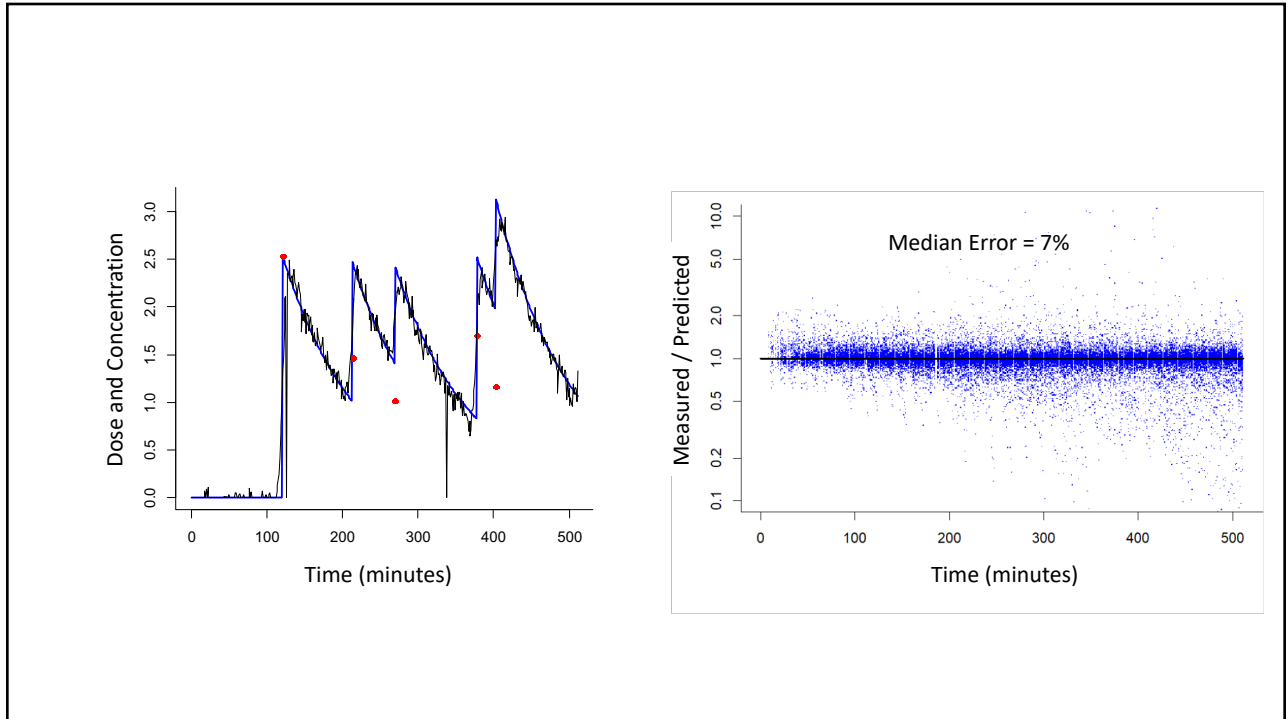
PK/PD Modeling with CNN

- Time invariance

- Repeatedly present the same input/output, just shifted to the right
 - e.g., same data, but starting in cell 1, then cell 2, then cell 3

- Superposition

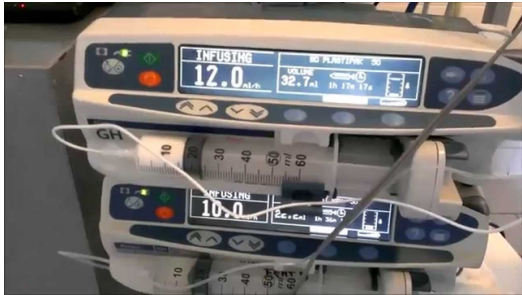

- Repeatedly present the same input/output, just scaled by x
 - e.g., same data, but 2x dose → 2x concentration



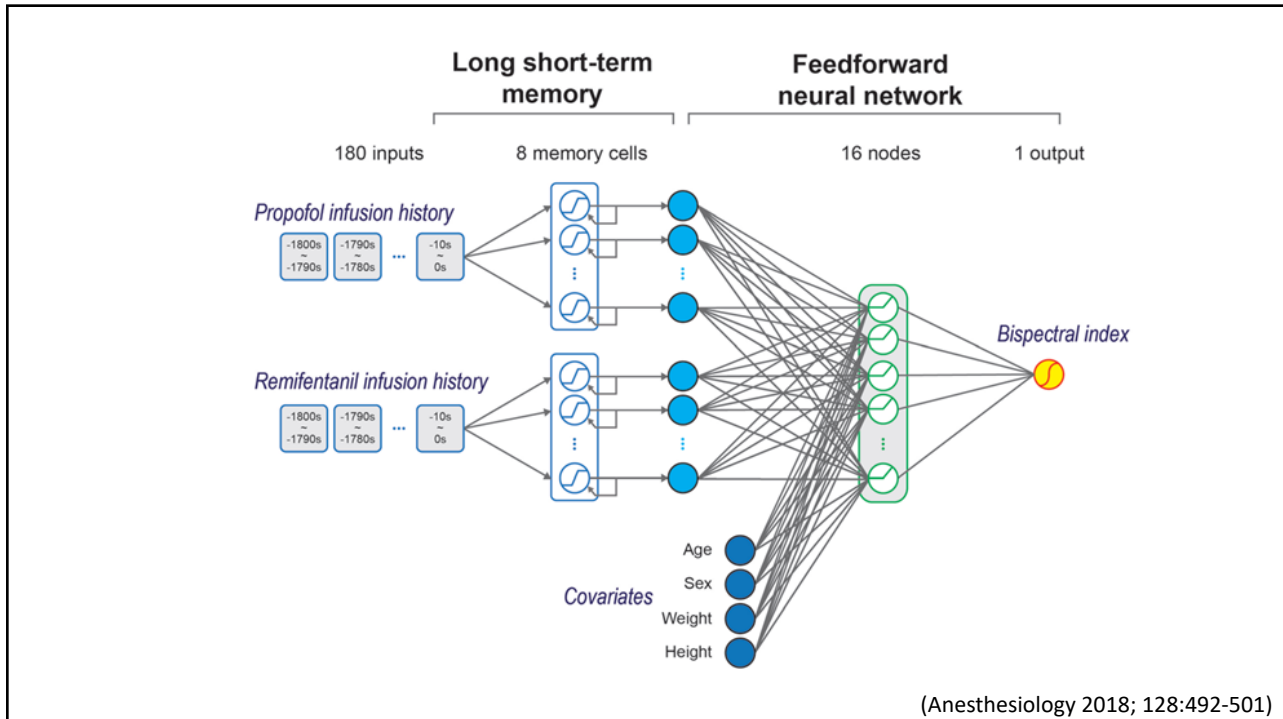
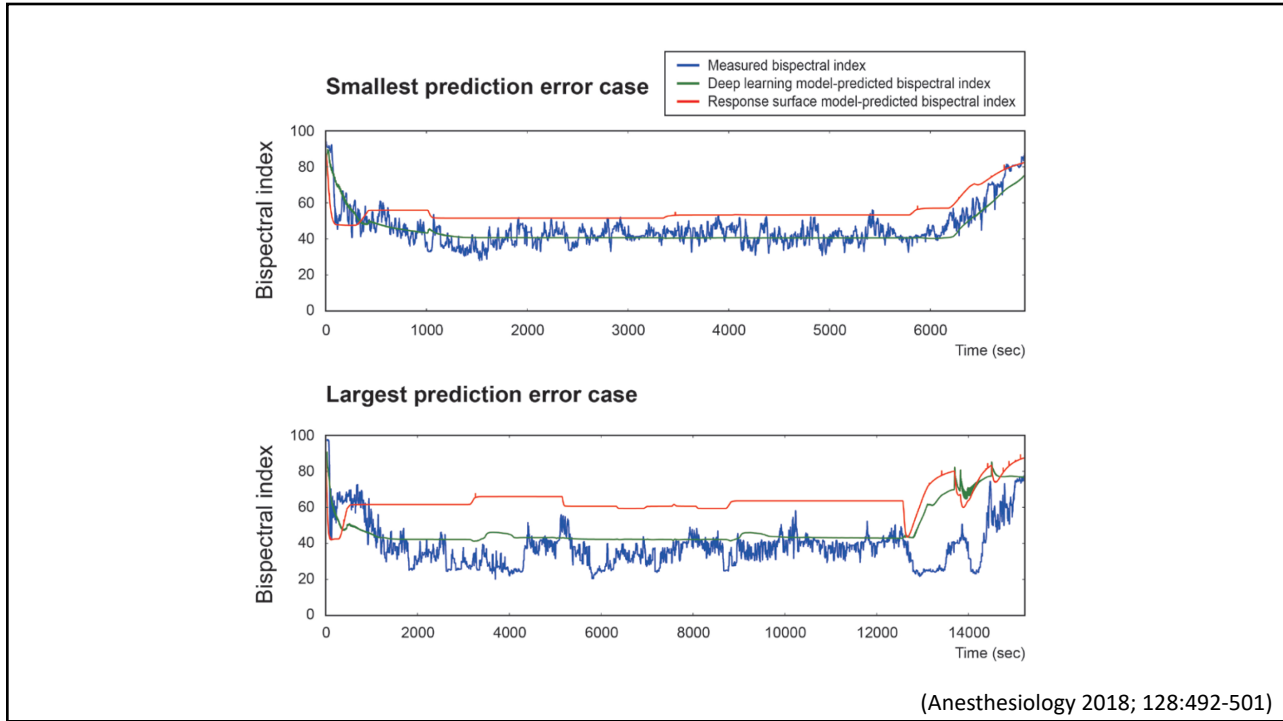
Prediction of Bispectral Index during Target-controlled Infusion of Propofol and Remifentanyl

A Deep Learning Approach

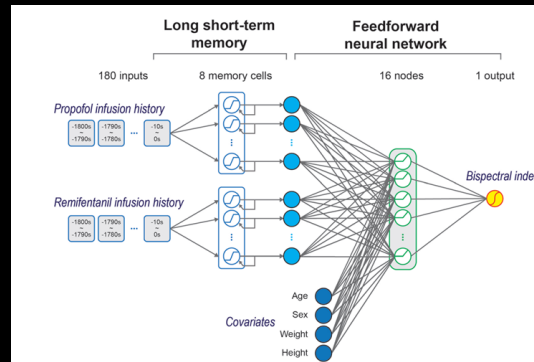
Hyung-Chul Lee, M.D., Ho-Geol Ryu, M.D., Ph.D., Eun-Jin Chung, M.D., Chul-Woo Jung, M.D., Ph.D.

(Anesthesiology 2018; 128:492-501)



Propofol_NN = LSTM(8, input=Propofol_Dose)
 Remifentaniil_NN = LSTM(8, input=Remifentaniil_Dose)
 Combined_NN = merge([Propofol_NN, Remifentaniil_NN, Covariates])
 NN2 = Dense(16, input=Combined_NN)
 BIS = Dense(1, input=NN2)



The Good

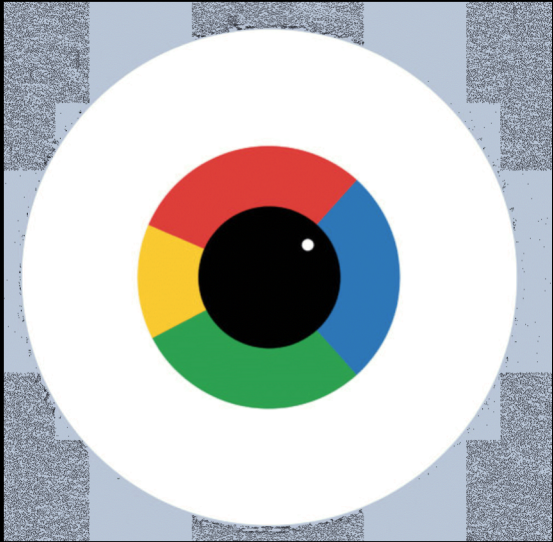
- Huge interest in many domains.
- They work.
 - If you aren't convinced, ask Alexa
- They always work.
- They scale.
 - Google Translate, built on a 7 layer LSTM model can translate any language to any other language.
 - Google uses neural networks to recognize places, faces, and features in millions of photographs every day
- TensorFlow + GPU/TPU = tractable Neural Networks

The Bad

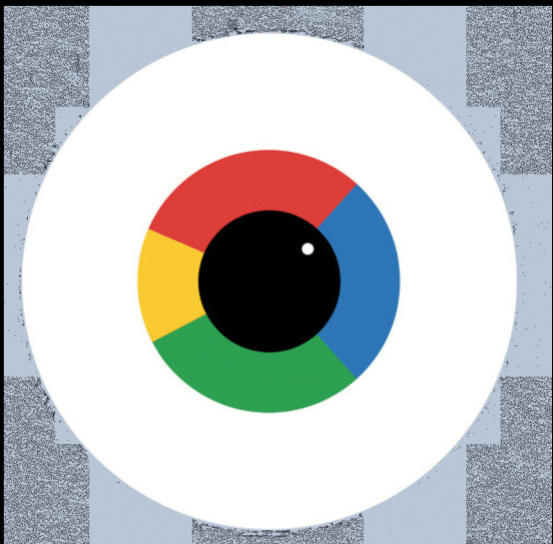
- They don't "know" anything.
 - Exceptionally good at predicting drug response
 - Probably can't explain why some patients differ from others.
- They can't extrapolate.
 - Can't see beyond the training data.
- Unclear how to incorporate prior knowledge and expertise.
- Almost nothing published about neural networks in clinical pharmacology.

The Ugly

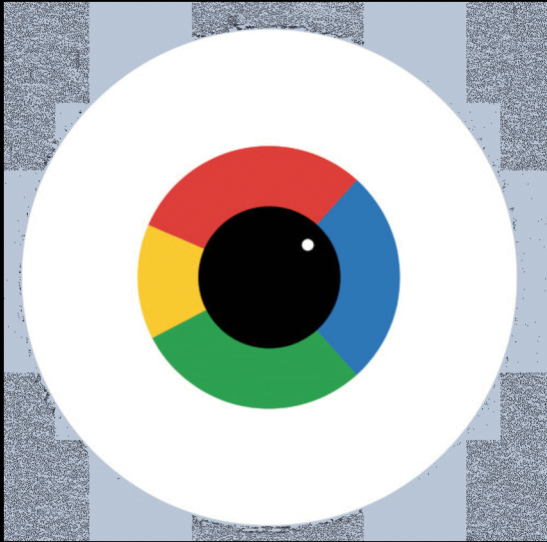
- They can quickly acquire superhuman ability in well defined systems.
- They are now being used to design neural networks.
- "Humans Need Not Apply"



Thank
You



Sleep
Well



Google Is
Watching
You