

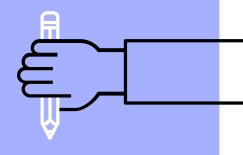
AGENDA











L. BACKGROUND



BACKGROUND

DATA:

Fashion-MNIST - a dataset of Zalando's article images—consisting of a training set of 60,000 examples and a test set of 10,000 examples

GOAL:

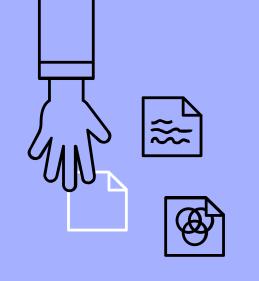
Build the best possible model to predict the class of the image as measured on the test set

TEAMWORK:

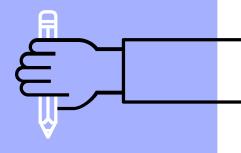




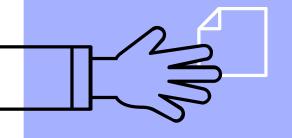








BUILD MODEL



Baseline Model - Architecture

Normalize **Units Units** Units **Flatten** = 300= 100= 10X_test **Activation** inputs **Activation** Activation Shape = 'relu' = 'softmax' = 'relu' **INPUTS OUTPUTS** HIDDEN1 HIDDEN2 **FLATTEN**

Baseline Model - Compile and fit model

- → Loss = 'sparse_categorical_crossentropy'
- → Optimizer=SGD

Learning rate = 0.01

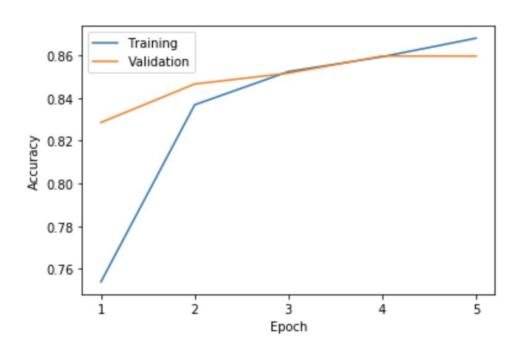
Momentum=0.9

- → Training dataset: X_train[:50000]
- → Validation dataset: X_train[50000:]
- **→ Batch_size** = 200
- \rightarrow Epochs = 5

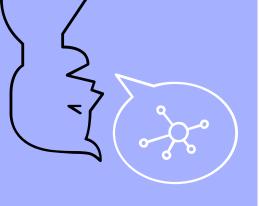


Baseline Model - Result

Accuray for test set = 0.8512









3. OPTIMIZATION



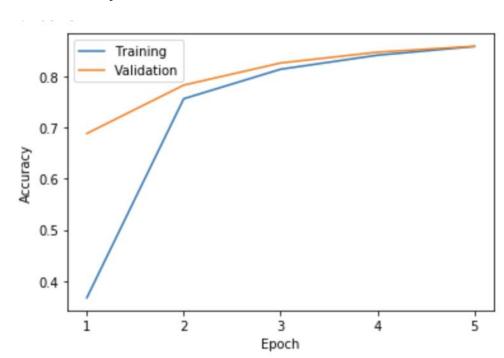
OPTIMIZATION - Add autoencoder

Non-trainable params: 0

Model: "model_1"		
Layer (type)	Output Shape	Param #
input_3 (InputLayer)	[(None, 28, 28)]	0
encoder (Model)	(None, 3, 3, 64)	23296
decoder (Model)	(None, 28, 28)	23233
flatten (Flatten)	(None, 784)	0
dense (Dense)	(None, 300)	235500
dense_1 (Dense)	(None, 100)	30100
out (Dense)	(None, 10)	1010
Total params: 313,139 Trainable params: 313,139		

Add autoencoder - Result

Accuray for test set = 0.8575





OPTIMIZATION

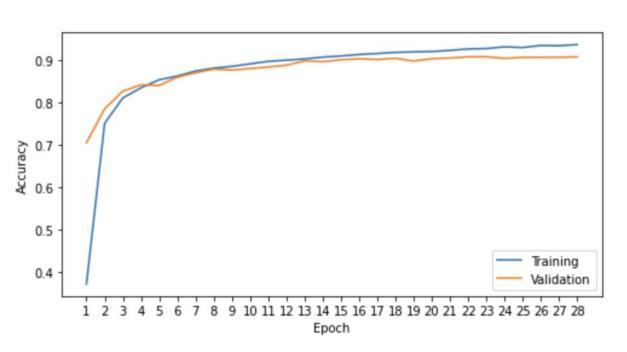
- Increase epoch and add early stop
- → Based on the autoencoder architecture
- → Increase epoch:

→ Add early stopping:

Patience = 5

Increase epoch and add early stop - Result

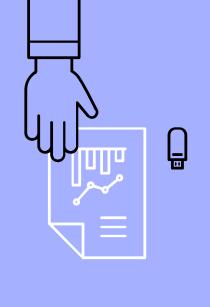
Accuray for test set = 90.33%





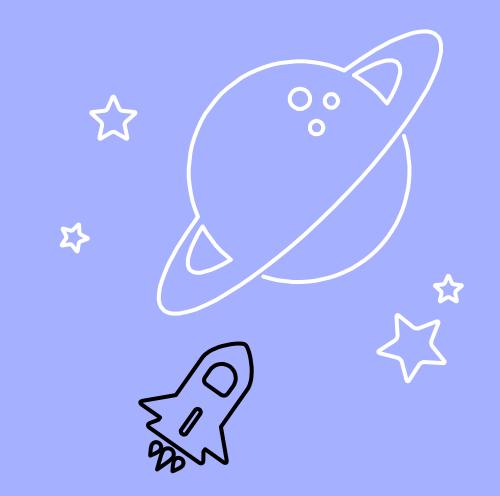
Accuracy Comparison

Models	Accuracy on test set
Baseline Model	0.8512
Add Conv2D layers	too slow to run
Add autoencoder	0.8575
Increase epoch and add early stop	0.9033





4. SUMMARY



Summary

Conv2D layers

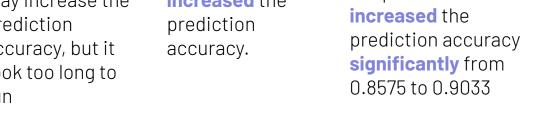
Adding Conv2D layers and increasing filters may increase the prediction accuracy, but it took too long to run

Autoencoder

Adding autoencoder with Conv2D slightly increased the

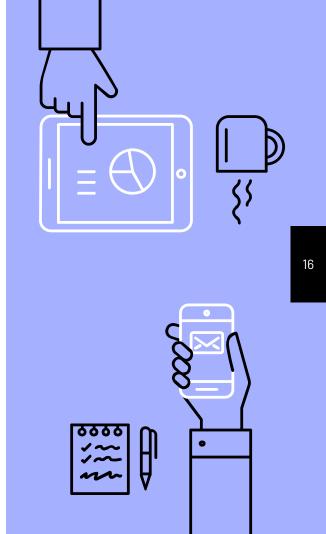
Increase Epoch with Early Stopping

Increasing epochs with early stopping to the previous model increased the prediction accuracy significantly from 0.8575 to 0.9033



Optimal Model

We got the optimal model with autoencoder and Conv2D and early stopping (at epoch 23), achieving 90% accuracy on test set.





THANKS!



Any questions?

Credit: Presentation template by <u>SlidesCarnival</u> Photographs by <u>Unsplash</u>

```
####### 2) Add Conv2D layers ######
# Create architecture
inputs = tf.keras.layers.Input(shape=(28,28,1), name='input')
# Conv2D layer
x = tf.keras.layers.Conv2D(filters=64, kernel_size=3, strides=1, padding="same",
activation="relu")(inputs)
x = tf.keras.layers.Conv2D(filters=64, kernel_size=3, strides=1, padding="same",
activation="relu")(inputs)
x = tf.keras.layers.MaxPooling2D(pool_size=2, strides=2, padding="valid")(x)
x = tf.keras.layers.Conv2D(filters=128, kernel_size=3, strides=1, padding="same",
activation="relu")(x)
x = tf.keras.layers.Conv2D(filters=128, kernel_size=3, strides=1, padding="same",
activation="relu")(x)
x = tf.keras.layers.MaxPooling2D(pool_size=2, strides=2, padding="valid")(x)
x = tf.keras.layers.Conv2D(filters=256, kernel_size=3, strides=1, padding="same",
activation="relu")(x)
x = tf.keras.layers.Conv2D(filters=256, kernel_size=3, strides=1, padding="same",
activation="relu")(x)
x = tf.keras.layers.MaxPooling2D(pool_size = 2, strides = 2, padding = "valid")(x)
x = tf.keras.layers.Flatten()(x)
x = tf.keras.layers.Dense(500, activation = 'relu')(x)
x = tf.keras.layers.Dense(250, activation = 'relu')(x)
outputs = tf.keras.layers.Dense(10, activation='softmax', name='out')(x)
model = tf.keras.Model(inputs=inputs, outputs=outputs)
print('Add Conv2D layers')
model.summary()
```