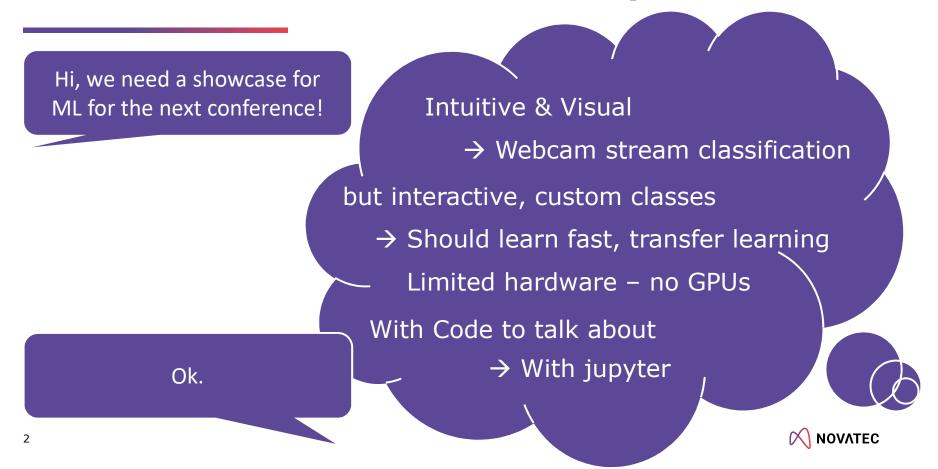
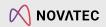


What do we want to achieve today?



Transfer Learning Example





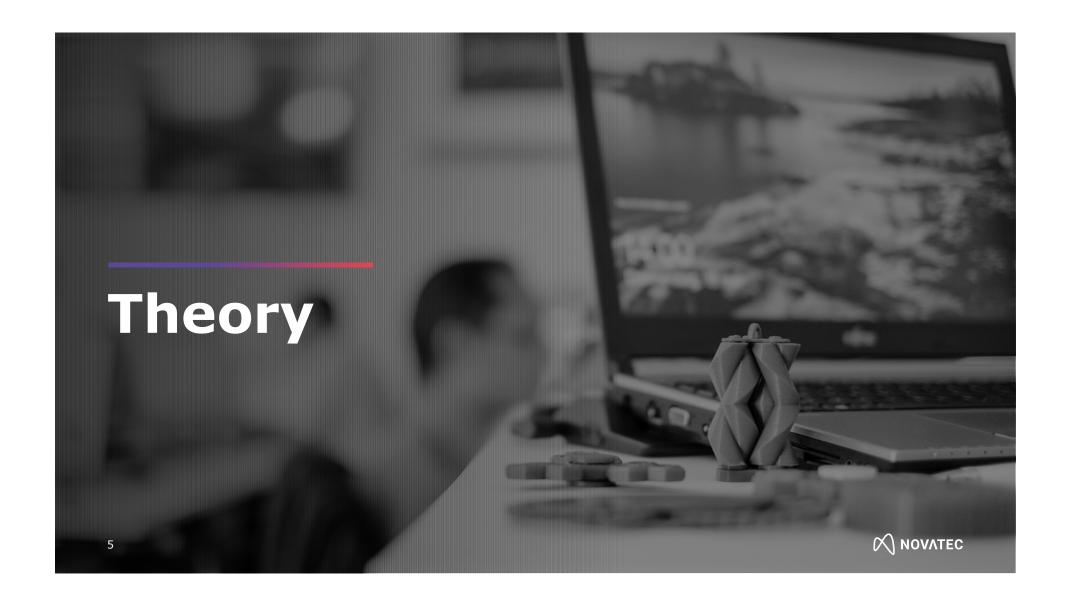
What we do differently

- Python
- Arbitrary number of classes
- Notebook style
 - less polished...
 - ... but visible code / structure
 - hackable
- Self contained
 - local deployment

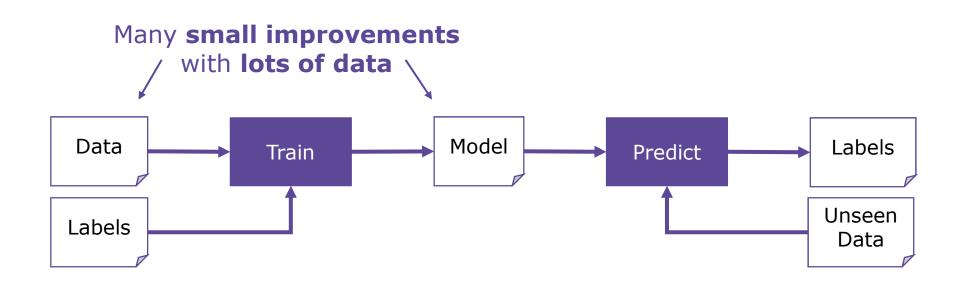


I see "Harald @ PyConDe" and I'm 100.00% certain about it



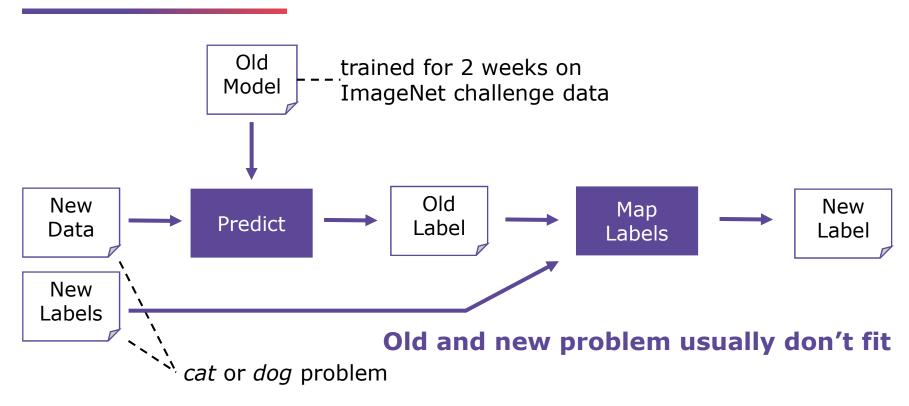


Standard Machine Learning Pipeline

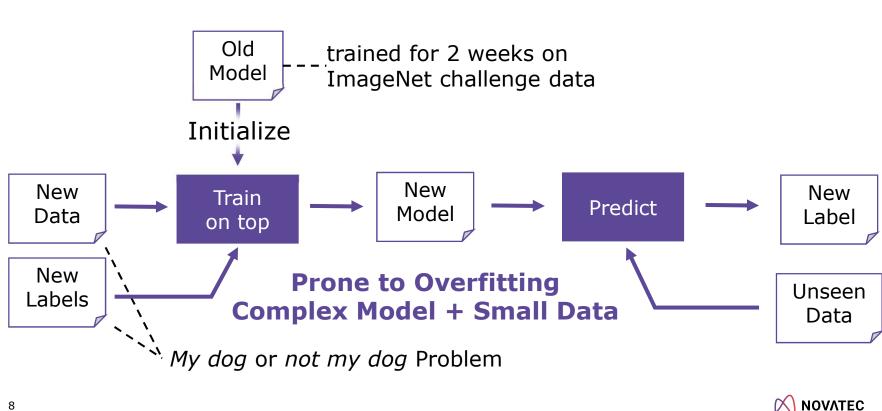




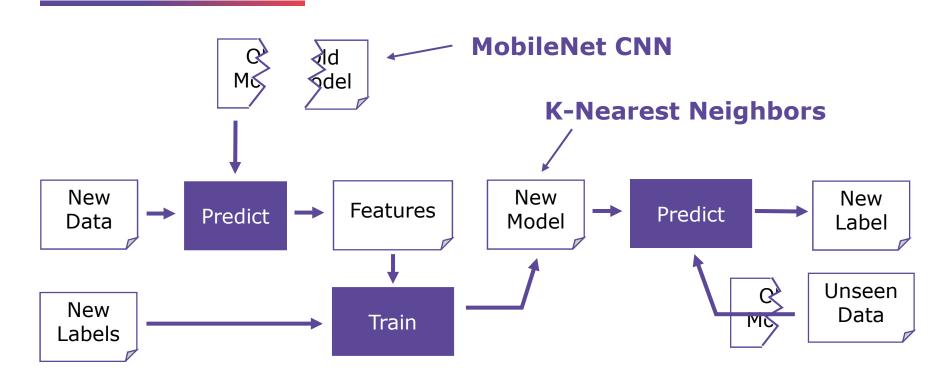
Transfer Learning Pipeline - Pretrained Model



Transfer Learning Pipeline – Fine-Tune Model

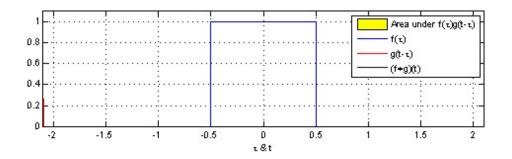


Transfer Learning Pipeline – Feature Extractor



Convolutional Neutral Networks (CNN)

- Developed by LeCun et al. 1998
- Why convolutional?
 - Because they use a **convolution** instead of a general matrix multiplication.
- What is a convolution?
 - "Filters" in image processing
 - The sum of the element-wise product between two matrices, in the discrete case



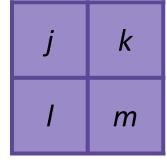
Convolution Operation

*

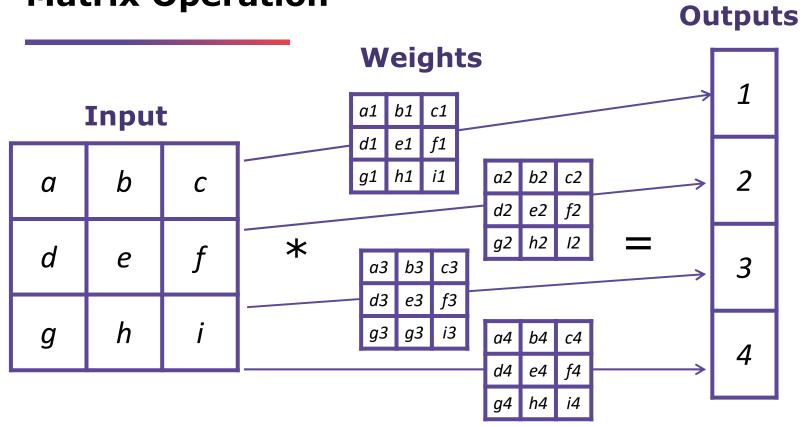
Input

а	b	С
d	e	f
g	h	i

Filter



Matrix Operation



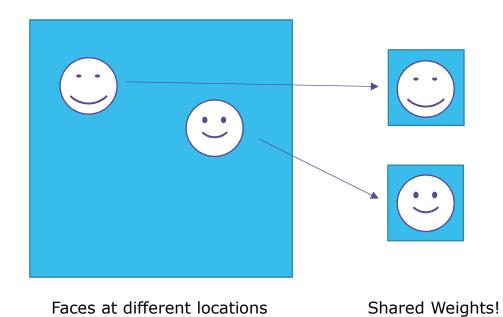
Motivation for CNNs - Neighborhood





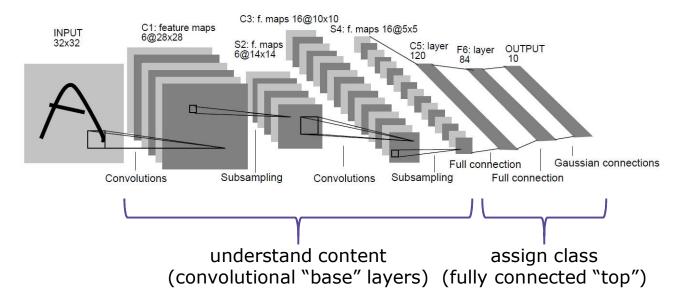


Motivation for CNN - Location Independent



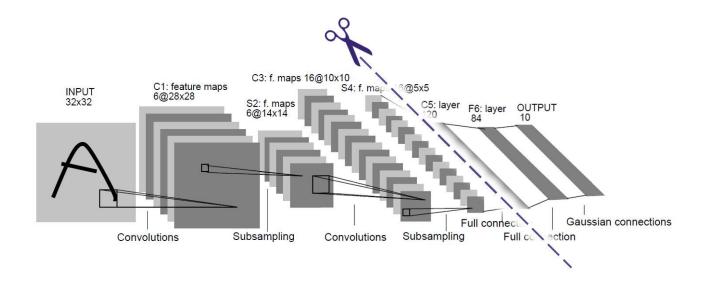


CNN Structure Example LeNet-5



Source: LeCun et al., 1998 NOVATEC

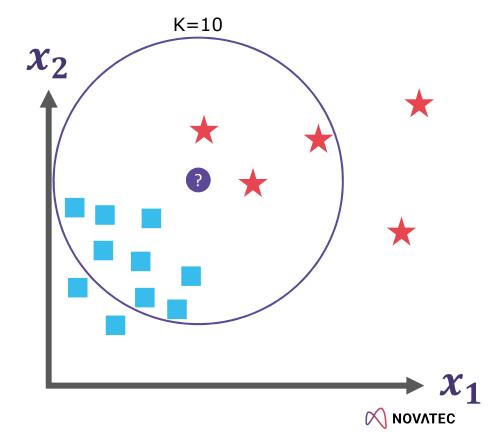
CNN as Feature Extractors

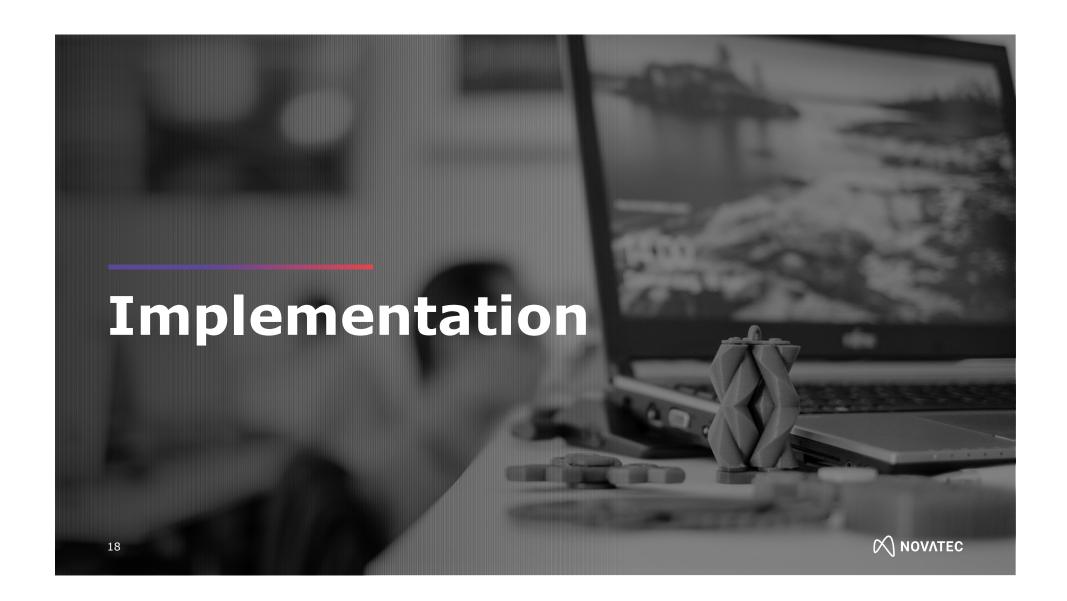


Source: <u>LeCun et al., 1998</u>

K Nearest Neighbors

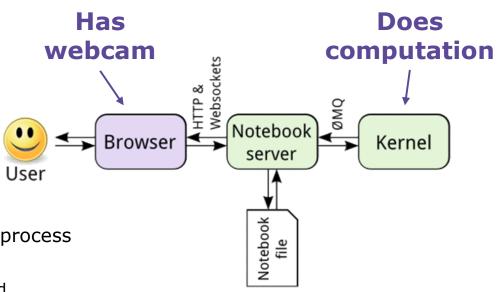
- Most simple classifier
- Training:
 - Memorize all labeled examples
- Inference:
 - Find k nearest examples
 - Make a majority vote





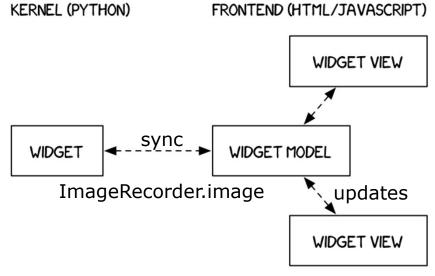
Notebook Style

- Why Jupyter?
 - Can easily be deployed
 - via docker and browser
 - remote and local alike
 - One place for code and story (mainly for collegues)
 - Cell structure is great to explain process
 - Be flexible
 - Show intermediate results if wanted
 - Do ad hoc changes



IPyWebRTC

- Used to access webcam image from Python
- Based upon ipywidgets
 - syncronized widget model
 - widgets to play video, audio, ...
- Additionally
 - Exposes <u>MediaStream API</u> and <u>WebRTC</u>
 - Wrapper for webcam media stream and media recorders
 - Allows to send content to peers



ImageRecorder.recording = true



Obtaining the Webcam content

- Create MediaStream from webcam
 - Constraints are passed on to [1]
- Create ImageRecorder from webcam stream
- Record image (omitted)
- PIL can translate PNG to RGBA
- BytesIO enables file like (i.e. random) access to the byte content

```
camera = CameraStream(
    constraints = {'video': {'width': 224, 'height': 224 }}
image_recorder = ImageRecorder(stream=camera)
img = PIL.Image.open(io.BytesIO(image recorder.image.value))
```

Load Pretrained Model

- Keras' applications provides easy access to pretrained networks
 - Load model without classification layers (include_top = False)
- Pretrained on ImageNet Challenge Data
 - 1.2 Mio images
 - 1000 classes
- Input Shape
 - 224 x 224 pixel, 3 RGB channels
 - default values for this model
- Pooling
 - Reduce feature vector
 - Take only the average pixel value for each channel

```
cnn = applications.mobilenet.MobileNet(
    include_top = False,
    weights = 'imagenet',
    input_shape = (224, 224, 3),
    pooling = 'avg',
)
```



Extract Image Features

- Obtain raw values from image
- Remove alpha channel
- Appy same preprocessing as in the original training data
 - Depending on backend, basically scaling to [-1;1]
- Wrap into array and "predict" the features
 - Keras, Scikit-Learn etc. usually work with <u>data sets</u> not individual instances
 - (224,224,3) \rightarrow (1,224,224,3)

```
def extract_features(img):
    img_array = np.asarray(img, dtype='int')
    img_wo_alpha = img_array[:, :, :3]
    img_wo_alpha = mobilenet.preprocess_input(img_wo_alpha)
    return cnn.predict(np.array([img_wo_alpha]))
```



Callbacks for fast data collection

- As a widget, ImageRecorder's properties are observable <u>Traitlets</u>
- This allows to register a callback on each value change.
- Each value change
 - Extracts the features from the change event
 - Stores features as input and label as desired output in global lists for training data

```
def training_callback(change):
    img = PIL.Image.open(io.BytesIO(change.new))
    x = extract_features(img)
    y = label
    X = np.append(X, x, axis=0)
    Y = np.append(Y, y)

training_recorder = ImageRecorder( stream = camera )
training_recorder.image.observe(training_callback, 'value')
```



Train Classifier

- scikit-learn's implementation of KNN
 - (or any other of their classifiers!)
 - Create Model
 - Fit Model
 - Done!

```
classification_model = neighbors.KNeighborsClassifier()
classification_model.fit(X, Y)
```



Use classifier

- Redo same steps as in the data collection
 - obtain image, preprocess, extract features
- Predict the most probable label
 - Returns array of size #images
- Get the probabilities
 - Returns array of size #images x #classes

```
def classify(image):
    feature_vector = extract_features(image)
    prediction = classification_model.predict(feature_vector)
    prediction_proba = classification_model.predict_proba(feature_vector)
    return prediction[0], np.max(prediction_proba)
```



Link everything with callbacks

- Output widget
 - HTML label
- Callback
 - uses the classifier
 - updates the output
 - Programmatically takes another image

```
img = PIL.Image.open(io.BytesIO(change.new))
    result, conf = classify(img)
    resultLabel.value = "{} - {}%".format(result, conf)
    if (proceed):
        inference_recorder.recording = True

recorder = ImageRecorder(stream=camera)
    recorder.image.observe(inference_callback, 'value')

widgets.VBox([recorder, resultLabel])
```

resultLabel = widgets.HTML()

def inference callback(change):

- Thus triggering another callback
- Combine ImageRecorder and output label e.g. via a vertical layout





