## Classification of Cars with Machine Learning

```
# This code installs and sets up the fastbook library as a dependency.
! [ -e /content ] && pip install -Uqq fastbook
import fastbook
fastbook.setup_book()
                                                 - 719.8/719.8 kB 11.8 MB/s eta 0:00:00
                                                   - 1.3/1.3 MB 22.6 MB/s eta 0:00:00
                                                     7.0/7.0 MB 75.1 MB/s eta 0:00:00
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                                                   - 110.5/110.5 kB 9.4 MB/s eta 0:00:00
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                                                  - 212.2/212.2 kB 9.2 MB/s eta 0:00:00
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                                                  - 114.2/114.2 kB 1.1 MB/s eta 0:00:00
                                                  - 158.8/158.8 kB <mark>8.6 MB/s</mark> eta 0:00:00
                                                   - 269.3/269.3 kB 7.0 MB/s eta 0:00:00
                                                    - 1.6/1.6 MB 39.1 MB/s eta 0:00:00
     Mounted at /content/gdrive
# Import dependencies
from fastbook import *
from fastai.vision.widgets import *
# Setting key
key = os.environ.get('AZURE_SEARCH_KEY', '74f7dde4575e4799800cae9dbb7d6e6f')
# Search Images
search_images_bing
     <function fastbook.search_images_bing(key, term, min_sz=128, max_images=150)>
# Search image of real cars
results = search_images_bing(key, 'carros')
ims = results.attrgot('contentUrl')
len(ims)
     150
# Getting image from yellow car
ims = ['https://images.cdn.circlesix.co/image/1/640/0/uploads/articles/dsc_2049_60809-564c6f63611b8.jpg']
# Save the image
dest = 'images/yellow_car.jpg'
download_url(ims[0], dest)
                                              -819200.00% [8192/-1 00:00<00:00]
                                              -1638400.00% [16384/-1 00:00<00:00]
                                              -2457600.00% [24576/-1 00:00<00:00]
                                              -3276800.00% [32768/-1 00:00<00:00]
                                              -4096000.00% [40960/-1 00:00<00:00]
                                              -4915200.00% [49152/-1 00:00<00:00]
                                              -5734400.00% [57344/-1 00:00<00:00]
                                              -6553600.00% [65536/-1 00:00<00:00]
                                              -7372800.00% [73728/-1 00:00<00:00]
     Path('images/yellow_car.jpg')
# Open and show the image
im = Image.open(dest)
im.to_thumb(256,256)
```



```
# Setting avaliables car colors
cars_colors = 'black','white','blue', 'yellow','green'
path = Path('carros')
# Get image from each cars_colors and download it
if not path.exists():
        path.mkdir()
        for o in cars_colors:
                dest = (path/o)
                dest.mkdir(exist_ok=True)
                results = search_images_bing(key, f'{0} carros')
                download_images(dest, urls=results.attrgot('contentUrl'))
# Gets all the image files within the specified directory path and assigns them to the variable fns
fns = get_image_files(path)
fns
           (#705) [Path('carros/black/987f4963-b807-45fe-824a-a04e6e9b9eab.jpg'),Path('carros/black/ebe664e3-77d6-4981-9c99-
           5af1d893568c.jpg'),Path('carros/black/7baaf5c0-9b25-445d-a8ad-54710156cf48.jpg'),Path('carros/black/8dc4cd2e-13af-434c-b5ec-
           28ad5ef3e5f7.jpg'), Path('carros/black/600e316f-9119-42a6-8b37-f58bd7af7315.jpg'), Path('carros/black/67575ac7-c84d-4d58-827d-62458-827d-62458-62458-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-624666-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-62466-6246
           0280569edd67.jpg')...]
# Prepare the data for training a model on images of cars
cars = DataBlock(
        blocks=(ImageBlock, CategoryBlock),
        get_items=get_image_files,
        splitter=RandomSplitter(valid_pct=0.2, seed=42),
        get_y=parent_label,
        item_tfms=Resize(128))
# This code creates a dataloader object dls from the cars DataBlock using the path specified earlier as the source of the data.
dls = cars.dataloaders(path)
# Will display a batch of images from the validation set of the dataloaders object created from the cars DataBlock.
dls.valid.show batch()
```







# This code block resizes the images in the cars datablock to a fixed size of 128 pixels using the Resize transformation with ResizeMeth # creates a new dataloader with the resized images, and displays a batch of 4 unique images from the validation set with a maximum of 1

cars = cars.new(item\_tfms=Resize(128, ResizeMethod.Squish)) dls = cars.dataloaders(path) dls.valid.show\_batch(max\_n=4, nrows=1, unique=True)









# Resizes the images to have a height and width of 64 pixels using padding to maintain the aspect ratio of the images # and sets the padding mode to 'zeros'. Then it creates a new dataloader object based on the modified dataset and shows a batch of four

cars = cars.new(item\_tfms=Resize(64, ResizeMethod.Pad, pad\_mode='zeros')) dls = cars.dataloaders(path) dls.valid.show\_batch(max\_n=5, nrows=3)













# This code randomly crops images in the training set using RandomResizedCrop with a minimum scale of 0.89 and displays a batch of 4 uni

cars = cars.new(item\_tfms=RandomResizedCrop(224, min\_scale=0.89))
dls = cars.dataloaders(path)
dls.train.show\_batch(max\_n=10, nrows=2, unique=True)

```
blue blue blue blue blue blue blue

blue blue blue blue blue

blue blue blue blue blue
```

# This code sets up a convolutional neural network using the ResNet18 architecture
# and fine-tunes it on the given dataloaders for 4 epochs. The performance of the model is evaluated using the error\_rate metric.
learn = cnn\_learner(dls, resnet18, metrics=error\_rate)
learn.fine\_tune(4)

/usr/local/lib/python3.9/dist-packages/fastai/vision/learner.py:288: UserWarning: warn("`cnn\_learner` has been renamed to `vision\_learner` -- please update your c /usr/local/lib/python3.9/dist-packages/torchvision/models/\_utils.py:208: UserWarni warnings.warn(

/usr/local/lib/python3.9/dist-packages/torchvision/models/\_utils.py:223: UserWarni
warnings.warn(msg)

Downloading: "https://download.pytorch.org/models/resnet18-f37072fd.pth" to /root/ 100% 44.7M/44.7M [00:00<00:00, 206MB/s]

epoch	train_loss	valid_loss	error_rate	time
0	2.506444	0.805031	0.294964	00:54
epoch	train_loss	valid_loss	error_rate	time
0	0.780268	0.438975	0.107914	01:00
1	0.488997	0.343556	0.079137	00:53
2	0.330993	0.346030	0.064748	00:55
3	0.242773	0.353371	0.071942	00:52

#·The·code·creates·a·convolutional·neural·network·learner·object·learn·using·a·ResNet34·model·architecture·and·the·dataloaders·dls.

 $\# \cdot \text{It} \cdot \text{then} \cdot \text{fine-tunes} \cdot \text{the} \cdot \text{pre-trained} \cdot \text{model} \cdot \text{for} \cdot 4 \cdot \text{epochs} \cdot \text{with} \cdot \text{a} \cdot \text{base} \cdot \text{learning} \cdot \text{rate} \cdot \text{of} \cdot 0.1, \cdot \text{using} \cdot \text{the} \cdot \text{fine\_tune}() \cdot \text{method} \cdot \text{The} \cdot \text{metric} \cdot \text{used} \cdot \text{for} \cdot 1 \cdot \text{method} \cdot \text{method} \cdot \text{for} \cdot 1 \cdot \text{method} \cdot \text{method} \cdot 1 \cdot \text{method$ 

 $learn \cdot = \cdot cnn\_learner(dls, \cdot resnet34, \cdot metrics = error\_rate) \\ learn.fine\_tune(4, \cdot base\_lr = \cdot 0.1)$ 

/usr/local/lib/python3.9/dist-packages/torchvision/models/\_utils.py:223: UserWarni
warnings.warn(msg)

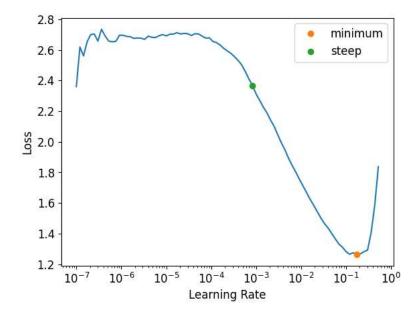
Downloading: "https://download.pytorch.org/models/resnet34-b627a593.pth" to /root/ 100% 83.3M/83.3M [00:00<00:00, 95.9MB/s]

epoch	train_loss	valid_loss	error_rate	time
0	1.236874	6.496347	0.366906	00:54
epoch	train_loss	valid_loss	error_rate	time
0	1.116350	1469.571777	0.820144	00:52
1	1.178550	76.311096	0.388489	00:53
2	0.907831	20.838860	0.503597	00:53
3	0.742382	1.238460	0.143885	00:52

# The code defines a cnn\_learner object with a ResNet34 architecture and the error\_rate metric.

# it then calls the fine\_tune method to train the model for 4 epochs with a default learning rate.

learn = cnn\_learner(dls, resnet34, metrics=error\_rate)
md = learn.lr\_find(suggest\_funcs=(minimum, steep))



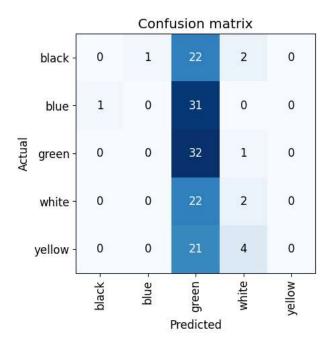
## # Print

print(f"Minimum/10: {md.minimum:.2e}, steepest point: {md.steep:.2e}")

Minimum/10: 1.74e-02, steepest point: 8.32e-04

- # creates an instance of the ClassificationInterpretation class from the cnn\_learner model
- # and uses it to plot a confusion matrix, which is a visual representation of the performance of the model in classifying the validation
- # showing the number of true positive, false positive, true negative, and false negative predictions for each class.

interp = ClassificationInterpretation.from\_learner(learn)
interp.plot\_confusion\_matrix()



- # generates a plot of the top losses in the validation set, showing the image with highest loss,
- # the predicted label and the true label. In this case, the plot shows the top 3 losses, with all images in the same row.

interp.plot\_top\_losses(3, nrows=1)

## Prediction/Actual/Loss/Probability green/black / 19.26 / 1.0green/black / 16.81 / 0.9green/blue / 16.62 / 1.00

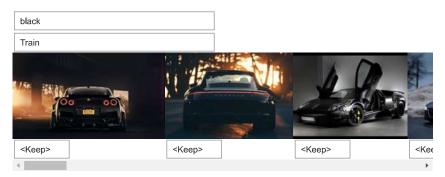






# Creates an instance of the ImageClassifierCleaner class with the specified model.

cleaner = ImageClassifierCleaner(learn)
cleaner



## ▼ Deploy

```
# Export
learn.export()
# Export the file
path = Path()
path.ls(file_exts='.pkl')
     (#1) [Path('export.pkl')]
# Load model
learn_inf = load_learner(path/'export.pkl')
# Chance
learn_inf.predict('images/yellow_car.jpg')
     ('white',
      tensor([2.8685e-05, 3.8698e-03, 1.6642e-01, 8.2601e-01, 3.6729e-03]))
# Options
learn_inf.dls.vocab
     ['black', 'blue', 'green', 'white', 'yellow']
# The code creates a button to upload an image file, a button to classify the image, an output widget to display the image
\sharp and a label widget to show the prediction and probability of the classification.
# The on_click_classify function is called when the classify button is clicked, which reads the uploaded image
# displays it in the output widget, and performs a prediction using the trained model, which is then displayed in the label widget.
btn_upload = widgets.FileUpload()
btn_run = widgets.Button(description='Classify')
out_pl = widgets.Output()
lbl_pred = widgets.Label()
def on_click_classify(change):
    img = PILImage.create(btn_upload.data[-1])
    out_pl.clear_output()
    with out_pl: display(img.to_thumb(128,128))
    pred,pred_idx,probs = model_inf.predict(img)
    lbl_pred.value = f'Prediction: {pred}; Probability: {probs[pred_idx]:.04f}'
```

```
btn_run.on_click(on_click_classify)

# Fake the image

btn_upload = SimpleNamespace(data = ['images/yellow_car.jpg'])

# Create

img = PILImage.create(btn_upload.data[-1])

# out_pl is an Output widget that can be used to display output from other widgets.
# Here, the code is clearing the output and displaying the thumbnail of an image of size 256x256 in the out_pl widget.
# However, since there is no image variable defined in this code, it will throw an error.

out_pl = widgets.Output()
out_pl.clear_output()
with out_pl: display(img.to_thumb(256,256))
out_pl
```



btn\_upload, btn\_run, out\_pl, lbl\_pred])

```
# This line of code uses the predict method of a learner object to make predictions on an input image
pred,pred_idx,probs = learn_inf.predict(img)
# The code creates a label widget and sets its value to a string containing the predicted label and probability for an image classificat
lbl_pred = widgets.Label()
lbl_pred.value = f'Prediction: {pred}; Probability: {probs[pred_idx]:.04f}'
lbl_pred
     Prediction: white; Probability: 0.8260
# Run the classify
btn_run = widgets.Button(description='Classify')
btn_run
            Classify
# Classify function
def on_click_classify(change):
    img = PILImage.create(btn_upload.data[-1])
    out_pl.clear_output()
    with out_pl: display(img.to_thumb(128,128))
    pred,pred idx,probs = learn inf.predict(img)
    lbl_pred.value = f'Prediction: {pred}; Probability: {probs[pred_idx]:.04f}'
btn_run.on_click(on_click_classify)
# Widget of file upload
btn_upload = widgets.FileUpload()
# Widget showing the chance
VBox([widgets.Label('Select the car from your computer'),
```

Select the car from your computer

Upload (0)

Classify



from google.colab import drive
drive.mount('/content/drive')

Mounted at /content/drive

✓ 3s conclusão: 16:57