```
#hide
! [ -e /content ] && pip install -Uqq fastbook
import fastbook
fastbook.setup_book()
                                                 - 719.8/719.8 KB 13.2 MB/s eta 0:00:00
                                                - 462.8/462.8 KB 48.5 MB/s eta 0:00:00
                                                  - 6.3/6.3 MB 21.4 MB/s eta 0:00:00
                                                   • 1.3/1.3 MB 76.8 MB/s eta 0:00:00
                                                - 213.0/213.0 KB 28.0 MB/s eta 0:00:00
                                                - 132.0/132.0 KB 18.5 MB/s eta 0:00:00
                                                - 190.3/190.3 KB 25.3 MB/s eta 0:00:00
                                                  - 7.6/7.6 MB 112.4 MB/s eta 0:00:00
                                                   1.6/1.6 MB 18.8 MB/s eta 0:00:00
                                                - 140.6/140.6 KB 18.8 MB/s eta 0:00:00
     Mounted at /content/gdrive
#hide
from fastbook import *
from fastai.vision.widgets import *
```

# → From Model to Production

# ▼ The Practice of Deep Learning

Starting Your Project

▼ The State of Deep Learning

Computer vision

Text (natural language processing)

Combining text and images

Tabular data

Recommendation systems

Other data types

The Drivetrain Approach

# **Gathering Data**

## - clean

To download images with Bing Image Search, sign up at <u>Microsoft Azure</u> for a free account. You will be given a key, which you can copy and enter in a cell as follows (replacing 'XXX' with your key and executing it):

```
len(ims)
     150
#hide
ims = ['https://hips.hearstapps.com/hmg-prod/images/2023-mclaren-artura-101-1655218102.jpg?crop=1.00xw:0.847xh;0,0.153xh&resize=1200:*']
dest = 'images/car.jpg'
download_url(ims[0], dest)
                                              105.27% [106496/101168 00:00<00:00]
     Path('images/car.jpg')
im = Image.open(dest)
im.to_thumb(128,128)
```



```
car_colors = 'blue', 'red', 'purple', 'green'
path = Path('cars')
if not path.exists():
    path.mkdir()
    for o in car_colors:
        dest = (path/o)
        dest.mkdir(exist_ok=True)
        results = search images bing(key, f'{o} car')
        download_images(dest, urls=results.attrgot('contentUrl'))
fns = get_image_files(path)
     (#586) [Path('cars/purple/859201f1-ce1a-400f-ab39-64d6e0df11a3.jpg'),Path('cars/purple/ced0696c-b50b-4431-93fc-
     3484883c71c9.jpg'),Path('cars/purple/f2a77fbf-9245-47d2-bb57-743beb82919c.jpg'),Path('cars/purple/01f3ba65-cbc1-4eb4-a3ae-
     acfdb3da1e37.jpg'),Path('cars/purple/cb95c5d4-ed04-490d-afab-1bc57b63571e.jpg'),Path('cars/purple/4d83988b-b719-4f43-9fec-
     ff7fdbf45729.jpg'),Path('cars/purple/a84ca3a6-5f76-42d5-9ab8-60bd89d57d66.jpg'),Path('cars/purple/68037e93-0fbb-4c49-a4ff-
     0b49f9f92117.jpg'),Path('cars/purple/b0bec7ca-96e7-4f34-8879-7b77638d1b7e.jpg'),Path('cars/purple/55a17b16-d4ba-4ab4-acf3-
     028b70eced24.jpg')...]
failed = verify_images(fns)
failed
     (#12) [Path('cars/purple/71312cb0-107d-483e-863b-f02a1b52e8ac.jpg'),Path('cars/purple/bcbf1b76-61ff-44b1-8ed0-
     da3fcf60f214.jpg'),Path('cars/purple/3546d899-a093-4caf-b1b2-ae0771fec461.jpg'),Path('cars/blue/e655a507-57df-44e5-9594-
     0fe9271cb552.jpg'),Path('cars/blue/ca508949-3317-4cec-8ec9-0c82c7c9eb8d.jpg'),Path('cars/blue/b91c2b66-2d54-474b-b211-
     a663ec6759f6.jpg'),Path('cars/blue/c449d093-4d63-40fa-b8a5-2412ed99f9ee.jpg'),Path('cars/red/fa9e41a4-1c57-4bdc-a60f-
     7f2d8939b5b8.jpg'),Path('cars/red/255e6385-d53f-49d9-86ca-4d409c0a8e2c.jpg'),Path('cars/green/c6b37647-4e33-4060-87a0-
     5494a7fe3728.jpg')...]
failed.map(Path.unlink);
```

Sidebar: Getting Help in Jupyter Notebooks

End sidebar

## ▼ From Data to DataLoaders

```
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))
dls = cars.dataloaders(path)
dls.valid.show batch(max n=6, nrows=1)
```













cars = cars.new(item\_tfms=Resize(128, ResizeMethod.Squish))
dls = cars.dataloaders(path)
dls.valid.show\_batch(max\_n=6, nrows=1)













cars = cars.new(item\_tfms=Resize(128, ResizeMethod.Pad, pad\_mode='zeros'))
dls = cars.dataloaders(path)
dls.valid.show\_batch(max\_n=6, nrows=1)













cars = cars.new(item\_tfms=RandomResizedCrop(128, min\_scale=0.3))
dls = cars.dataloaders(path)
dls.train.show\_batch(max\_n=6, nrows=1, unique=True)













## ▼ Data Augmentation

cars = cars.new(item\_tfms=Resize(128), batch\_tfms=aug\_transforms(mult=2))
dls = cars.dataloaders(path)
dls.train.show\_batch(max\_n=6, nrows=2, unique=True)









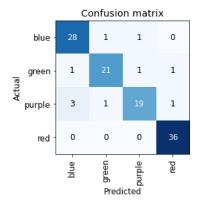




# ▼ Training Your Model, and Using It to Clean Your Data

```
cars = cars.new(
    item_tfms=RandomResizedCrop(224, min_scale=0.5),
    batch_tfms=aug_transforms())
dls = cars.dataloaders(path)
learn = vision_learner(dls, resnet18, metrics=error_rate)
learn.fine_tune(4)
     /usr/local/lib/python3.8/dist-packages/torchvision/models/_utils.py:208: UserWarni
     /usr/local/lib/python3.8/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, 119MB/s]
      epoch train_loss valid_loss error_rate time
                                       0.491228 00:33
          0
               2.089920
                           1.097556
      epoch train_loss valid_loss
                                     error_rate time
               0.745895
                           0.473360
                                       0.149123 00:33
               0.548345
                                       0.087719 00:41
          1
                           0.267099
          2
               0.417467
                           0.226167
                                       0.070175 00:37
          3
               0.324407
                           0.221799
                                       0.087719 00:32
```

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



interp.plot\_top\_losses(6, nrows=2)

#### Prediction/Actual/Loss/Probability

red/green / 6.40 / 0.93 blue/purple / 4.65 / 0.99 red/purple / 2.49 / 0.57

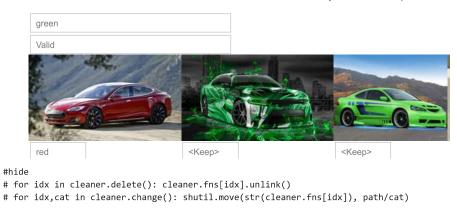
blue/purple / 1.63 / 0.73purple/green / 1.50 / 0.73green/purple / 1.45 / 0.76







cleaner = ImageClassifierCleaner(learn)
cleaner



## Turning Your Model into an Online Application

### ▼ Using the Model for Inference

```
learn.export()

path = Path()
path.ls(file_exts='.pkl')
    (#1) [Path('export.pkl')]

learn_inf = load_learner(path/'export.pkl')

learn_inf.predict('images/car.jpg')
    ('blue',
        TensorBase(0),
        TensorBase([9.9995e-01, 1.7602e-05, 1.7173e-06, 3.4695e-05]))

learn_inf.dls.vocab
    ['blue', 'green', 'purple', 'red']
```

### Creating a Notebook App from the Model

lbl\_pred = widgets.Label()

lbl\_pred

lbl\_pred.value = f'Prediction: {pred}; Probability: {probs[pred\_idx]:.04f}'

Prediction: blue; Probability: 0.9179

```
btn_run = widgets.Button(description='Classify')
btn_run
            Classify
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)
#hide
#Putting back btn_upload to a widget for next cell
btn_upload = widgets.FileUpload()
VBox([widgets.Label('Select your car!'),
      btn_upload, btn_run, out_pl, lbl_pred])
     Select your car!
           Upload (1)
            Classify
     Prediction: blue: Probability: 0.9179
```

▼ Turning Your Notebook into a Real App

```
#hide
# !pip install voila
# !jupyter serverextension enable --sys-prefix voila
```

Deploying your app

### How to Avoid Disaster

Unforeseen Consequences and Feedback Loops

Get Writing!

## Questionnaire

- 1. Provide an example of where the bear classification model might work poorly in production, due to structural or style differences in the training data.
- 2. Where do text models currently have a major deficiency?
- 3. What are possible negative societal implications of text generation models?
- 4. In situations where a model might make mistakes, and those mistakes could be harmful, what is a good alternative to automating a process?
- 5. What kind of tabular data is deep learning particularly good at?
- 6. What's a key downside of directly using a deep learning model for recommendation systems?
- 7. What are the steps of the Drivetrain Approach?
- 8. How do the steps of the Drivetrain Approach map to a recommendation system?
- 9. Create an image recognition model using data you curate, and deploy it on the web.

- 10. What is DataLoaders?
- 11. What four things do we need to tell fastai to create DataLoaders?
- 12. What does the splitter parameter to DataBlock do?
- 13. How do we ensure a random split always gives the same validation set?
- 14. What letters are often used to signify the independent and dependent variables?
- 15. What's the difference between the crop, pad, and squish resize approaches? When might you choose one over the others?
- 16. What is data augmentation? Why is it needed?
- 17. What is the difference between item\_tfms and batch\_tfms?
- 18. What is a confusion matrix?
- 19. What does export save?
- 20. What is it called when we use a model for getting predictions, instead of training?
- 21. What are IPython widgets?
- 22. When might you want to use CPU for deployment? When might GPU be better?
- 23. What are the downsides of deploying your app to a server, instead of to a client (or edge) device such as a phone or PC?
- 24. What are three examples of problems that could occur when rolling out a bear warning system in practice?
- 25. What is "out-of-domain data"?
- 26. What is "domain shift"?
- 27. What are the three steps in the deployment process?

#### ▼ Further Research

- 1. Consider how the Drivetrain Approach maps to a project or problem you're interested in.
- 2. When might it be best to avoid certain types of data augmentation?
- 3. For a project you're interested in applying deep learning to, consider the thought experiment "What would happen if it went really, really well?"
- 4. Start a blog, and write your first blog post. For instance, write about what you think deep learning might be useful for in a domain you're interested in.