

```

In [2]: import numpy as np
import pandas as pd
import joblib
import matplotlib inline
import matplotlib.pyplot as plt
import matplotlib.pyplot as plt

In [3]: import tensorflow as tf
from tensorflow import keras

In [4]: from sklearn.datasets import fetch_california_housing
housing = fetch_california_housing()

In [5]: print(housing.feature_names)

['MedInc', 'HouseAge', 'AveRooms', 'AveBedrms', 'Population', 'AveOccup', 'Latitude', 'Longitude']

In [6]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train_full, y_test = train_test_split(housing.data, housing.target, random_state = 42)
X_train, X_valid, y_train, y_valid = train_test_split(X_train_full, y_train_full, random_state = 42)

In [7]: from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_valid = scaler.transform(X_valid)
X_test = scaler.transform(X_test)

In [8]: np.random.seed(42)
tf.random.set_seed(42)

In [9]: X_train.shape

Out[9]: (11619, 8)

In [10]: model = keras.models.Sequential([
keras.layers.Dense(30, activation = "relu", input_shape=[8]),
keras.layers.Dense(30, activation = "relu"),
keras.layers.Dense(1)
])

In [11]: model.compile(loss = "mean_squared_error",
optimizer = keras.optimizers.SGD(lr=1e-3),
metrics = ['mae'])

In [12]: model.summary()

Model: "sequential"
Layer (type) Output Shape Param #
=====
dense (Dense) (None, 30) 270
dense_1 (Dense) (None, 30) 930
dense_2 (Dense) (None, 1) 31
Total params: 1,231
Trainable params: 1,231
Non-trainable params: 0

In [13]: model_history = model.fit(X_train, y_train, epochs = 20, validation_data = (X_valid, y_valid))

Epoch 1/20
363/363 [=====] - loss: 1.8906 - mae: 0.9990 - val_loss: 0.7126 - val_mae: 0.6368
Epoch 2/20
363/363 [=====] - loss: 0.4577 - mae: 0.4642 - val_loss: 0.6880 - val_mae: 0.5784
Epoch 3/20
363/363 [=====] - loss: 0.5934 - mae: 0.5618 - val_loss: 0.5803 - val_mae: 0.5352
Epoch 4/20
363/363 [=====] - loss: 0.5557 - mae: 0.5398 - val_loss: 0.5186 - val_mae: 0.5297
Epoch 5/20
363/363 [=====] - loss: 0.5272 - mae: 0.5237 - val_loss: 0.4895 - val_mae: 0.5022
Epoch 6/20
363/363 [=====] - loss: 0.5833 - mae: 0.5113 - val_loss: 0.4951 - val_mae: 0.4934
Epoch 7/20
363/363 [=====] - loss: 0.4854 - mae: 0.5010 - val_loss: 0.4861 - val_mae: 0.4838
Epoch 8/20
363/363 [=====] - loss: 0.4769 - mae: 0.4924 - val_loss: 0.4554 - val_mae: 0.4753
Epoch 9/20
363/363 [=====] - loss: 0.4578 - mae: 0.4857 - val_loss: 0.4413 - val_mae: 0.4671
Epoch 10/20
363/363 [=====] - loss: 0.4474 - mae: 0.4797 - val_loss: 0.4379 - val_mae: 0.4623
Epoch 11/20
363/363 [=====] - loss: 0.4393 - mae: 0.4744 - val_loss: 0.4396 - val_mae: 0.4638
Epoch 12/20
363/363 [=====] - loss: 0.4318 - mae: 0.4783 - val_loss: 0.4587 - val_mae: 0.4573
Epoch 13/20
363/363 [=====] - loss: 0.4261 - mae: 0.4674 - val_loss: 0.3997 - val_mae: 0.4517
Epoch 14/20
363/363 [=====] - loss: 0.4202 - mae: 0.4636 - val_loss: 0.3956 - val_mae: 0.4497
Epoch 15/20
363/363 [=====] - loss: 0.4155 - mae: 0.4613 - val_loss: 0.3916 - val_mae: 0.4464
Epoch 16/20
363/363 [=====] - loss: 0.4112 - mae: 0.4591 - val_loss: 0.3937 - val_mae: 0.4445

```

[illegible]

```
In [27]: y_pred = model.predict(X_new)
print(y_pred)
print(y_test[:3])

[[0.5328768]
 [1.991398]
 [0.46087]]
[0.477  0.458  5.00801]
```

Functional API

```
In [28]: del model

In [29]: keras.backend.clear_session()

In [20]: input_ = keras.layers.Input(shape=(X_train.shape[1:]))
hidden1 = keras.layers.Dense(30, activation="relu")(input_)
hidden2 = keras.layers.Dense(30, activation="relu")(hidden1)
concat = keras.layers.concatenate([input_, hidden2])
output = keras.layers.Dense(5)(concat)
model = keras.models.Model(inputs=[input_], outputs=[output])

In [21]: model.summary()
```

```

Model: "functional_1"
Layer (type)                   Output Shape          Param #   Connected to
-----
input_1 (InputLayer)          [(None, 8)]           0         (None, 8)
dense_1 (Dense)                (None, 38)            278        input_1[0][0]
dense_1_1 (Dense)              (None, 38)            938        dense[0][0]
concatenate (Concatenate)      (None, 38)            0         input_1[0][0]
                                                dense_1[0][0]
dense_2 (Dense)                (None, 1)             39         concatenate[0][0]
=====
Total params: 1,239
Trainable params: 4,239
Non-trainable params: 0
=====

In [22]: model.compile(loss = "mean_squared_error",
optimizer = keras.optimizers.SGD(lr=1e-3),
metrics = ["mae"])

In [28]: model_history = model.fit(X_train, y_train, epochs = 50, validation_data = (X_valid, y_valid))

Epoch 1/50
363/363 [=====] - 15 2es/step - loss: 0.3710 - mae: 0.4330 - val_loss: 0.3721 - val_mae: 0.4246
Epoch 2/50
363/363 [=====] - 15 2es/step - loss: 0.3700 - mae: 0.4326 - val_loss: 0.4372 - val_mae: 0.4316
Epoch 3/50
363/363 [=====] - 15 2es/step - loss: 0.3692 - mae: 0.4328 - val_loss: 0.4056 - val_mae: 0.4266

```

Epoch 4/20	15 2m52s/step	loss: 0.3880	val_loss: 0.4315	val_loss: 0.3777	val_mae: 0.4268
362/362					
Epoch 5/20	15 2m52s/step	loss: 0.3675	val_loss: 0.4310	val_loss: 0.3728	val_mae: 0.4246
362/362					
Epoch 6/20	15 2m52s/step	loss: 0.3663	val_loss: 0.4304	val_loss: 0.3662	val_mae: 0.4245
362/362					
Epoch 7/20	15 2m52s/step	loss: 0.3658	val_loss: 0.4297	val_loss: 0.3579	val_mae: 0.4236
362/362					
Epoch 8/20	15 2m52s/step	loss: 0.3649	val_loss: 0.4294	val_loss: 0.3579	val_mae: 0.4284
362/362					
Epoch 9/20	15 2m52s/step	loss: 0.3640	val_loss: 0.4287	val_loss: 0.3884	val_mae: 0.4221
362/362					
Epoch 10/20	15 2m52s/step	loss: 0.3633	val_loss: 0.4281	val_loss: 0.3942	val_mae: 0.4218
362/362					
Epoch 11/20	15 2m52s/step	loss: 0.3627	val_loss: 0.4273	val_loss: 0.3469	val_mae: 0.4183
362/362					
Epoch 12/20	15 2m52s/step	loss: 0.3617	val_loss: 0.4258	val_loss: 0.4210	val_mae: 0.4234
362/362					
Epoch 13/20	15 2m52s/step	loss: 0.3612	val_loss: 0.4264	val_loss: 0.3765	val_mae: 0.4222
362/362					
Epoch 14/20	15 2m52s/step	loss: 0.3604	val_loss: 0.4255	val_loss: 0.3525	val_mae: 0.4189
362/362					
Epoch 15/20	15 2m52s/step	loss: 0.3597	val_loss: 0.4254	val_loss: 0.3837	val_mae: 0.4212
362/362					
Epoch 16/20	15 2m52s/step	loss: 0.3590	val_loss: 0.4255	val_loss: 0.3434	val_mae: 0.4169
362/362					
Epoch 17/20	15 2m52s/step	loss: 0.3583	val_loss: 0.4244	val_loss: 0.3974	val_mae: 0.4204
362/362					
Epoch 18/20	15 2m52s/step	loss: 0.3581	val_loss: 0.4242	val_loss: 0.3741	val_mae: 0.4179
362/362					
Epoch 19/20	15 2m52s/step	loss: 0.3571	val_loss: 0.4233	val_loss: 0.3637	val_mae: 0.4171
362/362					
Epoch 20/20	15 2m52s/step	loss: 0.3566	val_loss: 0.4229	val_loss: 0.4175	val_mae: 0.4211
362/362					

```

363/363 [=====] - 1s 28s/step - loss: 0.3560 - val_loss: 0.4223 - val_f1: 0.3743 - val_mae: 0.4196
epoch 27/50
363/363 [=====] - 1s 28s/step - loss: 0.3552 - val_loss: 0.4274 - val_f1: 0.3436 - val_mae: 0.4226
epoch 28/50
363/363 [=====] - 1s 28s/step - loss: 0.3548 - val_loss: 0.4223 - val_f1: 0.3367 - val_mae: 0.4189
epoch 29/50
363/363 [=====] - 1s 28s/step - loss: 0.3540 - val_loss: 0.4288 - val_f1: 0.3773 - val_mae: 0.4162
epoch 30/50
363/363 [=====] - 1s 28s/step - loss: 0.3534 - val_loss: 0.4289 - val_f1: 0.3935 - val_mae: 0.4174
epoch 31/50
363/363 [=====] - 1s 28s/step - loss: 0.3530 - val_loss: 0.4295 - val_f1: 0.3899 - val_mae: 0.4164
epoch 32/50
363/363 [=====] - 1s 28s/step - loss: 0.3527 - val_loss: 0.4293 - val_f1: 0.3367 - val_mae: 0.4111
epoch 33/50
363/363 [=====] - 1s 28s/step - loss: 0.3520 - val_loss: 0.4194 - val_f1: 0.3893 - val_mae: 0.4172
epoch 34/50
363/363 [=====] - 1s 28s/step - loss: 0.3516 - val_loss: 0.4196 - val_f1: 0.3591 - val_mae: 0.4147
epoch 35/50
363/363 [=====] - 1s 28s/step - loss: 0.3508 - val_loss: 0.4189 - val_f1: 0.3871 - val_mae: 0.4146
epoch 36/50
363/363 [=====] - 1s 28s/step - loss: 0.3501 - val_loss: 0.4182 - val_f1: 0.3525 - val_mae: 0.4139
epoch 37/50
363/363 [=====] - 1s 28s/step - loss: 0.3497 - val_loss: 0.4180 - val_f1: 0.3920 - val_mae: 0.4172
epoch 38/50
363/363 [=====] - 1s 28s/step - loss: 0.3496 - val_loss: 0.4184 - val_f1: 0.3444 - val_mae: 0.4118
epoch 39/50
363/363 [=====] - 1s 28s/step - loss: 0.3487 - val_loss: 0.4175 - val_f1: 0.3494 - val_mae: 0.4126
epoch 40/50
363/363 [=====] - 1s 28s/step - loss: 0.3484 - val_loss: 0.4172 - val_f1: 0.3934 - val_mae: 0.4167
epoch 41/50
363/363 [=====] - 1s 28s/step - loss: 0.3478 - val_loss: 0.4172 - val_f1: 0.3322 - val_mae: 0.4083
epoch 42/50
363/363 [=====] - 1s 28s/step - loss: 0.3475 - val_loss: 0.4163 - val_f1: 0.3625 - val_mae: 0.4125
epoch 43/50
363/363 [=====] - 1s 28s/step - loss: 0.3470 - val_loss: 0.4165 - val_f1: 0.3396 - val_mae: 0.4094

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```
Epoch 39/50
362/362 [=====] - 12s 28s/step - loss: 0.3463 - mae: 0.4157 - val_loss: 0.3447 - val_mae: 0.4092
Epoch 40/50
362/362 [=====] - 12s 28s/step - loss: 0.3407 - mae: 0.4155 - val_loss: 0.3587 - val_mae: 0.4081
Epoch 41/50
362/362 [=====] - 12s 28s/step - loss: 0.3458 - mae: 0.4150 - val_loss: 0.3616 - val_mae: 0.4102
Epoch 42/50
362/362 [=====] - 12s 28s/step - loss: 0.3455 - mae: 0.4153 - val_loss: 0.3574 - val_mae: 0.4101
Epoch 43/50
362/362 [=====] - 12s 28s/step - loss: 0.3446 - mae: 0.4142 - val_loss: 0.3691 - val_mae: 0.4097
Epoch 44/50
362/362 [=====] - 12s 28s/step - loss: 0.3440 - mae: 0.4140 - val_loss: 0.3734 - val_mae: 0.4126
Epoch 45/50
362/362 [=====] - 12s 28s/step - loss: 0.3441 - mae: 0.4144 - val_loss: 0.3734 - val_mae: 0.4060
Epoch 46/50
362/362 [=====] - 12s 28s/step - loss: 0.3431 - mae: 0.4129 - val_loss: 0.3473 - val_mae: 0.4084
Epoch 47/50
362/362 [=====] - 12s 28s/step - loss: 0.3429 - mae: 0.4135 - val_loss: 0.3490 - val_mae: 0.4072
Epoch 48/50
362/362 [=====] - 12s 28s/step - loss: 0.3424 - mae: 0.4127 - val_loss: 0.3625 - val_mae: 0.4090
Epoch 49/50
362/362 [=====] - 12s 28s/step - loss: 0.3427 - mae: 0.4130 - val_loss: 0.3320 - val_mae: 0.4068
Epoch 50/50
362/362 [=====] - 12s 28s/step - loss: 0.3416 - mae: 0.4120 - val_loss: 0.3254 - val_mae: 0.4030

In [29]:
mae_test = model.evaluate(X_test, y_test)

162/162 [=====] - 0s 1ms/step - loss: 0.3421 - mae: 0.4117

In [30]:
model.history.history

Out[30]: {'loss': (0.376998864688073,
0.378837784131595)
```

0.3681808495189867,
0.3680393395185422,
0.3675465641127777,
0.3667607658305352,
0.36584290882085435,
0.3648883552895304,
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0.3617465954950405,
0.3612287475337982,
0.3607391518942527,
0.3596731272354889,
0.3589896951184887,
0.358288789475503567,
0.3581177899390751,
0.3571268894903821,
0.3565226296964605,
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0.3530381310461807,
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0.3490548893177386,
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0.3479649972931565,
0.34747474566071957

```

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0.3444152321699448,
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0.3420871417018077,
0.34237808612454224,
0.34265134417520075,
0.35105499746602071,
mac: [ 0.432927389918812134,
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0.432007117009705,
0.43153429611177207,
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0.423243611717241,
0.4232335841112599]

```

```

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0.4189318139794784,
0.4182458873725053,
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0.41830270038964239,
0.4176204602120865,
0.41724684166320984,
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0.414252121938274903,
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0.41724684166320984,
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0.41830270038964239,
0.4179545342922107,
0.4182458873725053,
0.4189318139794784,
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0.4181028229209928,
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0.4208518266678564,
0.4284718471385054,
0.4222621891307933"]

```

0.35790387146926579,
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0.380356077770901,
0.3598698298477173,
0.38796284295183386,
0.3525407612323761,
0.35205229546649593,
0.3446025848388672,
0.368427469474638,
0.3884080880854834,
0.232343669474638,
0.352527222986884,
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[illegible]

```

0.4173786705887789,
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0.40839899197390903,
0.407164909981947,
0.409796445356599,
0.4086229464798344,
0.403769403690602}]

In [3]: import pandas as pd
        pd.DataFrame(model.history.history).plot(figsize = (4,3))
        plt.grid(True)
        plt.gca().set_ylim(0,3)
        plt.show()
```



```

In [34]: model.save("my_func_model.h5")

In [35]: %pwd
Out[35]: 'C:\Users\h22de'

In [36]: del model

In [38]: keras.backend.clear_session()

In [39]: model = keras.models.load_model("my_func_model.h5")

In [40]: model.summary()

```

```

Model: "Functional_1"
Layer (type)                Output Shape          Param #   Connected to
input_1 (InputLayer)        [(None, 0)]           0         (None, 0)
dense (Dense)               (None, 38)            278        input_1[0][0]
dense_1 (Dense)             (None, 38)            938        dense[0][0]
concatenate (Concatenate)   (None, 38)            0         input_1[0][0]
                                                dense_1[0][0]
dense_2 (Dense)             (None, 1)             39         concatenate[0][0]
=====
Total params: 1,239
Trainable params: 1,239
Non-trainable params: 0

```

```

In [41]: y_pred = model.predict(X_new)
         print(y_pred)

[[0.78474243]
 [1.0626977]
 [4.186468 ]]

```

Using Callbacks during Training

```

In [42]: del model

```

```
[42]: keras.backend.clear_session()
      np.random.seed(42)
      tf.random.set_seed(42)

[43]: model = keras.models.Sequential([
      keras.layers.Dense(30, activation = "relu", input_shape=[8]),
      keras.layers.Dense(30, activation = "relu"),
      keras.layers.Dense(1)
      ])

[44]: model.compile(loss = "mean_squared_error",
      optimizer = keras.optimizers.SGD(lr=1e-3),
      metrics = ["mse"])

[45]: checkpoint_cb = keras.callbacks.ModelCheckpoint("model-epoch-020.h5")

[46]: history = model.fit(X_train, y_train, epochs = 10, validation_data = (X_valid, y_valid), callbacks = [checkpoint_cb])

[47]: Epoch 1/10
      300/300 [=====] - 0s 951us/step - loss: 1.8866 - mae: 0.9909 - val_loss: 0.7126 - val_mae: 0.6368
Epoch 2/10
      300/300 [=====] - 0s 851us/step - loss: 0.6577 - mae: 0.6642 - val_loss: 0.6808 - val_mae: 0.5784
Epoch 3/10
      300/300 [=====] - 0s 851us/step - loss: 0.5934 - mae: 0.5619 - val_loss: 0.5803 - val_mae: 0.5352
Epoch 4/10
      300/300 [=====] - 0s 867us/step - loss: 0.5557 - mae: 0.5398 - val_loss: 0.5166 - val_mae: 0.5207
Epoch 5/10
```

```

Epoch 9/10 ..... -> 8s 848us/step - loss: 0.5272 - mae: 0.5237 - val_loss: 0.4895 - val_mae: 0.5822
Epoch 10/10 ..... -> 8s 858us/step - loss: 0.5032 - mae: 0.5113 - val_loss: 0.4951 - val_mae: 0.4934
Epoch 11/10 ..... -> 8s 821us/step - loss: 0.4854 - mae: 0.5010 - val_loss: 0.4861 - val_mae: 0.4838
Epoch 12/10 ..... -> 8s 678us/step - loss: 0.4709 - mae: 0.4924 - val_loss: 0.4554 - val_mae: 0.4753
Epoch 13/10 ..... -> 8s 693us/step - loss: 0.4578 - mae: 0.4857 - val_loss: 0.4413 - val_mae: 0.4671
Epoch 14/10 ..... -> 8s 731us/step - loss: 0.4474 - mae: 0.4797 - val_loss: 0.4379 - val_mae: 0.4623

[In [49]] del model
          keras.backend.clear_session()

[In [49]] model = keras.models.load_model("Model-10.h5")

[In [50]] mse_test = model.evaluate(X_test, y_test)

162/162 [.....] -> 8s 442us/step - loss: 0.4382 - mean_absolute_error: 0.427

Save Best Only

[In [51]] del model
          keras.backend.clear_session()

[In [52]] model = keras.models.Sequential([
            keras.layers.Dense(64, activation = 'relu', input_shape=(3)),

```

```

keras.layers.Dense(30, activation = "relu"),
keras.layers.Dense(1)
])

[In 53]: model.compile(loss = "mean_squared_error",
optimizer = keras.optimizers.SGD(lr=1e-3),
metrics = ["mse"])

[In 54]: checkpoint_cb = keras.callbacks.ModelCheckpoint("Best_Model.h5", save_best_only = True)

[In 56]: history = model.fit(X_train, y_train, epochs = 10, validation_data =(X_valid, y_valid), callbacks = [checkpoint_cb])

Epoch 1/10
361/361 [-----] 0s 10ms/step - loss: 0.4740 - mse: 0.2740
WARNING:tensorflow:ModelCheckpoint method on 'train_batch_end' is slow compared to the batch time (batch time: 0.0009s vs on_train_batch_end time: 0.0011s). Check your callbacks.
Epoch 2/10
361/361 [-----] 0s 10ms/step - loss: 0.4581 - mse: 0.2610
Epoch 3/10
361/361 [-----] 0s 10ms/step - loss: 0.4380 - mse: 0.2366
Epoch 4/10
361/361 [-----] 0s 10ms/step - loss: 0.4200 - mse: 0.2160
Epoch 5/10
361/361 [-----] 0s 10ms/step - loss: 0.4051 - mse: 0.2006
Epoch 6/10
361/361 [-----] 0s 10ms/step - loss: 0.3920 - mse: 0.1875
Epoch 7/10
361/361 [-----] 0s 10ms/step - loss: 0.3800 - mse: 0.1760
Epoch 8/10
361/361 [-----] 0s 10ms/step - loss: 0.3690 - mse: 0.1660
Epoch 9/10
361/361 [-----] 0s 10ms/step - loss: 0.3590 - mse: 0.1570
Epoch 10/10
361/361 [-----] 0s 10ms/step - loss: 0.3500 - mse: 0.1490
```

```

363/363 [=====] - 8s 786us/step - loss: 0.4738 - mae: 0.4944 - val_loss: 0.4359 - val_mae: 0.4723
Epoch 9/20
363/363 [=====] - 8s 781us/step - loss: 0.4563 - mae: 0.4856 - val_loss: 0.4267 - val_mae: 0.4608
Epoch 10/20
363/363 [=====] - 8s 680us/step - loss: 0.4431 - mae: 0.4790 - val_loss: 0.4224 - val_mae: 0.4593

In [57]:
model = keras.models.load_model("Best_Model.h5")
mse_test = model.evaluate(X_test, y_test)

162/162 [=====] - 0s 469us/step - loss: 0.4366 - mean_absolute_error: 0.4682

In [ ]:
```