```
In [2]:
```

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-pytho
# For example, here's several helpful packages to load
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files
under the input directory
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
   for filename in filenames:
       os.path.join(dirname, filename)
# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserve
d as output when you create a version using "Save & Run All"
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of
the current session
```

In [3]:

```
import fastai
import pandas as pd
from fastai import *
from fastai.vision import *
from fastai.vision.all import *
from fastai.data.all import *
```

In [4]:

```
csv_path = "/kaggle/input/skin-cancer-mnist-ham10000/HAM10000_metadata.csv"
df = pd.read_csv(csv_path)
df.sort_values(by="image_id")
```

Out[4]:

	lesion_id	image_id	dx	dx_type	age	sex	localization
4349	HAM_0000550	ISIC_0024306	nv	follow_up	45.0	male	trunk
4263	HAM_0003577	ISIC_0024307	nv	follow_up	50.0	male	lower extremity
4217	HAM_0001477	ISIC_0024308	nv	follow_up	55.0	female	trunk
3587	HAM_0000484	ISIC_0024309	nv	follow_up	40.0	male	trunk
1451	HAM_0003350	ISIC_0024310	mel	histo	60.0	male	chest
			•••				
1721	HAM_0004304	ISIC_0034316	mel	histo	85.0	male	upper extremity
1888	HAM_0006376	ISIC_0034317	mel	histo	70.0	female	lower extremity
121	HAM_0000344	ISIC_0034318	bkl	histo	55.0	male	trunk
7440	HAM_0000747	ISIC_0034319	nv	histo	30.0	male	trunk
7363	HAM_0002244	ISIC_0034320	nv	histo	25.0	female	chest

10015 rows × 7 columns

In [5]:

```
short_to_full_name_dict = {
   "akiec" : "Bowen's disease",
   "bcc" : "basal cell carcinoma" ,
```

```
"bkl" : "benign keratosis-like lesions",
   "df" : "dermatofibroma",
   "mel" : "melanoma",
   "nv" : "melanocytic nevi",
   "vasc" : "vascular lesions",
}

In [6]:

img_to_class_dict = df.loc[:, ["image_id", "dx"]]
img_to_class_dict = img_to_class_dict.to_dict('list')
img_to_class_dict = {img_id : short_to_full_name_dict[disease] for img_id,disease in zip
   (img_to_class_dict['image_id'], img_to_class_dict['dx']) }
```

```
Out[6]:
```

```
[('ISIC_0027419', 'benign keratosis-like lesions'), ('ISIC_0025030', 'benign keratosis-like lesions'), ('ISIC_0026769', 'benign keratosis-like lesions'), ('ISIC_0025661', 'benign keratosis-like lesions'), ('ISIC_0031633', 'benign keratosis-like lesions')]
```

[x for x in img to class dict.items()][:5]

In [7]:

```
def get_label_from_dict(path):
    return img_to_class_dict[path.stem]

skin_db = DataBlock(
    blocks=(ImageBlock, CategoryBlock),
    item_tfms=[Resize(450), DihedralItem()],
    batch_tfms=RandomResizedCrop(size=224, min_scale=0.75, max_scale=1.0),
    get_items=get_image_files,
    splitter=RandomSplitter(valid_pct=0.3, seed=42),
    get_y=get_label_from_dict,
)

img_path = "/kaggle/input/skin-cancer-mnist-ham10000"
dls = skin_db.dataloaders(img_path)
```

In [8]:

dls.show batch (max n=16, nrows=4)





melanocytic nevi







In [9]:

!pip install GPUtil

Collecting GPUtil Downloading GPUtil-1.4.0.tar.gz (5.5 kB)

Preparing metadata (setup.py) ... done

Building wheels for collected packages: GPUtil

Building wheel for GPUtil (setup.py) ... done

Created wheel for GPUtil: filename=GPUtil-1.4.0-py3-none-any.whl size=7409 sha256=e4163 1d11b86f26af02451cdb22760d0ea4514618eb32d8649277e09f884118a

Stored in directory: /root/.cache/pip/wheels/b1/e7/99/2b32600270cf23194c9860f029d3d5db0 75f250bc39028c045

Successfully built GPUtil

Installing collected packages: GPUtil

Successfully installed GPUtil-1.4.0

WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system package manager. It is recommended to use a virtual environment instead: https://pip.pypa.io/warnings/venv

In [10]:

```
from GPUtil import showUtilization as qpu usage
gpu usage()
import torch
torch.cuda.empty cache()
```

```
| ID | GPU | MEM |
 0 | 0% | 9% |
```

In [10]:

Initial Learner

learn = cnn learner(dls, resnet18, metrics=accuracy, opt_func=ranger) learn.fine_tune(epochs=30, freeze_epochs=3, base_lr=0.005, cbs=MixUp(0.5))

/opt/conda/lib/python3.7/site-packages/fastai/vision/learner.py:288: UserWarning: `cnn le arner` has been renamed to `vision learner` -- please update your code

warn("`cnn learner` has been renamed to `vision learner` -- please update your code") /opt/conda/lib/python3.7/site-packages/torchvision/models/ utils.py:209: UserWarning: The parameter 'pretrained' is deprecated since 0.13 and may be removed in the future, please use 'weights' instead.

f"The parameter '{pretrained_param}' is deprecated since 0.13 and may be removed in the future,

/opt/conda/lib/python3.7/site-packages/torchvision/models/ utils.py:223: UserWarning: Arg uments other than a weight enum or `None` for 'weights' are deprecated since 0.13 and may be removed in the future. The current behavior is equivalent to passing `weights=ResNet18 _Weights.IMAGENET1K_V1`. You can also use `weights=ResNet18_Weights.DEFAULT` to get the m ost up-to-date weights.

warnings.warn(msg)

Downloading: "https://download.pytorch.org/models/resnet18-f37072fd.pth" to /root/.cache/

epoch	train_loss	valid_loss	accuracy	time
0	2.043256	1.086300	0.691130	06:50
1	1.104947	0.624305	0.775503	04:13
2	0.885769	0.573451	0.796472	04:13
epoch	train_loss	valid_loss	accuracy	time
0	0.755308	0.468768	0.834249	04:26
1	0.707593	0.411693	0.858712	04:27
2	0.672564	0.375797	0.870528	04:27
3	0.650821	0.344235	0.889665	04:28
4	0.634279	0.345014	0.883342	05:43
5	0.601421	0.321498	0.891829	05:38
6	0.596203	0.397159	0.857880	04:40
7	0.585376	0.269682	0.914961	06:47
8	0.571816	0.366504	0.875021	06:23
9	0.552480	0.259081	0.913796	04:51
10	0.539393	0.275110	0.907805	04:28
11	0.522806	0.245165	0.914961	04:29
12	0.502613	0.199589	0.938426	04:29
13	0.487073	0.204677	0.935264	04:26
14	0.487707	0.238969	0.919288	04:27
15	0.453201	0.165753	0.950574	04:27
16	0.456821	0.166004	0.955899	04:26
17	0.443454	0.142854	0.964054	04:28
18	0.434803	0.138586	0.965385	04:23
19	0.407063	0.127556	0.966717	04:16
20	0.422691	0.122868	0.970544	04:17
21	0.416465	0.118928	0.971709	04:17
22	0.413294	0.119454	0.973207	04:17
23	0.401228	0.111835	0.971876	04:19
24	0.397174	0.107922	0.972708	04:19
25	0.389953	0.107366	0.975370	04:18
26	0.382036	0.104543	0.976868	04:17
27	0.395707	0.105295	0.976535	04:18
28	0.399156	0.102488	0.976202	04:18
29	0.385271	0.102930	0.976702	04:19

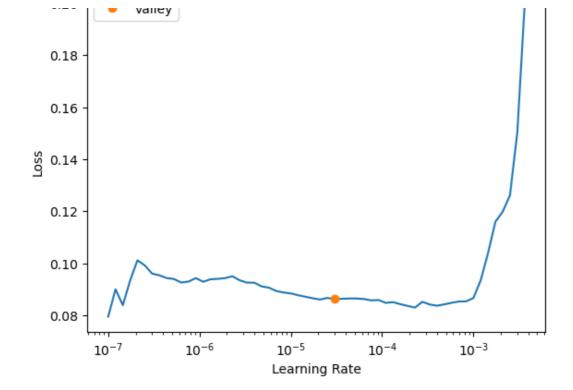
In [11]:

learn.lr_find()

Out[11]:

SuggestedLRs(valley=3.0199516913853586e-05)

0.20



In [11]:

learn = cnn_learner(dls, resnet18, metrics=accuracy, opt_func=ranger)
learn.fine_tune(epochs=100, freeze_epochs=3, base_lr=0.001, cbs=MixUp(0.5))

/opt/conda/lib/python3.7/site-packages/fastai/vision/learner.py:288: UserWarning: `cnn_le
arner` has been renamed to `vision_learner` -- please update your code
 warn("`cnn_learner` has been renamed to `vision_learner` -- please update your code")
/opt/conda/lib/python3.7/site-packages/torchvision/models/_utils.py:209: UserWarning: The
parameter 'pretrained' is deprecated since 0.13 and may be removed in the future, please
use 'weights' instead.

f"The parameter '{pretrained_param}' is deprecated since 0.13 and may be removed in the future, "

/opt/conda/lib/python3.7/site-packages/torchvision/models/_utils.py:223: UserWarning: Arg uments other than a weight enum or `None` for 'weights' are deprecated since 0.13 and may be removed in the future. The current behavior is equivalent to passing `weights=ResNet18 _Weights.IMAGENET1K_V1`. You can also use `weights=ResNet18_Weights.DEFAULT` to get the most up-to-date weights.

warnings.warn(msg)

Downloading: "https://download.pytorch.org/models/resnet18-f37072fd.pth" to /root/.cache/torch/hub/checkpoints/resnet18-f37072fd.pth

epoch	train_loss	valid_loss	accuracy	time
0	2.623863	1.633239	0.514062	07:50
1	1.757838	0.956417	0.717257	06:31
2	1.167635	0.647208	0.774338	04:50
epoch	train_loss	valid_loss	accuracy	time
0	1.029475	0.611293	0.786653	04:16
1	1.025281	0.573413	0.800466	04:15
2	0.972748	0.540607	0.810118	04:15
3	0.935108	0.515658	0.818938	04:16
4	0.913832	0.492599	0.826427	04:15
5	0.869049	0.467030	0.840406	04:16
6	0.828643	0.451438	0.839907	04:14
7	0.833526	0.430970	0.848394	04:15
8	0.786358	0.421381	0.854219	04:22

ерось	train loss 0.778362	valid Joss 0.400555	accuracy 0.861541	#ime 04:20
10	0.755470	0.389584	0.863704	04:20
11	0.722876	0.371343	0.871859	04:20
12	0.708535	0.356076	0.873856	04:20
13	0.699789	0.341446	0.880513	04:19
14	0.672050	0.329643	0.887668	04:21
15	0.648956	0.320767	0.892661	04:19
16	0.620975	0.312286	0.892495	04:18
17	0.632826	0.300594	0.899484	04:19
18	0.621343	0.282165	0.906640	04:22
19	0.598247	0.278068	0.905642	04:20
20	0.581979	0.273723	0.910468	04:17
21	0.573456	0.262675	0.915793	04:17
22	0.560739	0.254169	0.920785	04:19
23	0.544077	0.241437	0.923947	04:17
24	0.551939	0.236858	0.920453	04:19
25	0.529042	0.223149	0.928940	04:23
26	0.542143	0.223980	0.928940	04:24
27	0.533927	0.218993	0.932768	04:23
28	0.507415	0.210222	0.935763	04:27
29	0.515348	0.204135	0.940090	04:25
30	0.509696	0.206526	0.938093	04:24
31	0.485594	0.191996	0.941088	04:25
32	0.499555	0.192683	0.942087	04:24
33	0.482294	0.179857	0.946580	04:24
34	0.480651	0.180529	0.947579	04:24
35	0.472742	0.181830	0.946414	04:23
36	0.472239	0.175628	0.954568	04:17
37	0.463172	0.167395	0.958229	04:14
38	0.470043	0.159348	0.958063	04:14
39	0.465738	0.157575	0.958895	04:12
40	0.441938	0.157969	0.955899	04:12
41	0.453579	0.150187	0.961558	04:12
42	0.445276	0.154824	0.957730	04:12
43	0.452209	0.146467	0.959727	04:11
44	0.435502	0.145736	0.963888	04:13
45	0.436654	0.138192	0.965718	04:14
46	0.422257	0.143111	0.964054	04:12
47	0.441598	0.136049	0.967882	04:12
48	0.424899	0.135359	0.965219	04:16
49	0.423291	0.126672	0.970877	04:13
50	0.429966	0.133792	0.967882	04:16
51	0.424617	0.130365	0.969046	04:15
52	0.425676	0.130393	0.970211	04:15
	0.419661	0.126928	0.970045	04:16

ероср	train. 1955	valid loss 0.120545	30.970348	dime 04:15
55	0.415333	0.120936	0.973040	04:14
56	0.415825	0.121316	0.972708	04:15
57	0.416911	0.114889	0.973540	04:14
58	0.422089	0.120841	0.972375	04:12
59	0.415647	0.117358	0.974205	04:13
60	0.402800	0.112049	0.974871	04:14
61	0.406455	0.116414	0.972541	04:13
62	0.408684	0.112150	0.974871	04:12
63	0.406190	0.113503	0.971043	04:13
64	0.407260	0.113277	0.975537	04:13
65	0.410460	0.113543	0.974705	04:13
66	0.404603	0.113293	0.975204	04:11
67	0.396260	0.106829	0.976535	04:13
68	0.395441	0.106213	0.975037	04:13
69	0.401994	0.107519	0.975870	04:13
70	0.406793	0.106419	0.975703	04:15
71	0.399512	0.104255	0.978366	04:16
72	0.409145	0.102717	0.977367	04:13
73	0.391257	0.104020	0.976036	04:12
74	0.397598	0.106763	0.976369	04:12
75	0.379729	0.103312	0.976868	04:12
76	0.400708	0.104763	0.978033	04:12
77	0.397065	0.104964	0.975370	04:13
78	0.391041	0.105944	0.975370	04:14
79	0.410993	0.102685	0.975204	04:14
80	0.393323	0.100830	0.978033	04:11
81	0.380393	0.102779	0.976535	04:12
82	0.393615	0.103853	0.974871	04:14
83	0.404463	0.102744	0.975370	04:18
84	0.389813	0.104146	0.975204	04:18
85	0.389343	0.101350	0.976535	04:16
86	0.389991	0.099292	0.976036	04:14
87	0.372080	0.101788	0.976369	04:16
88	0.397181	0.101508	0.974705	04:15
89	0.384514	0.101313	0.977700	04:18
90	0.381983	0.101936	0.976868	04:12
91	0.386333	0.103397	0.976702	04:15
92	0.383298	0.101671	0.976369	04:15
93	0.374795	0.101256	0.977367	04:11
94	0.374678	0.101540	0.976868	04:13
95	0.395439	0.102006	0.977034	04:17
96	0.389326	0.100756	0.976702	04:16
97	0.391507	0.101753	0.976369	04:13
98	0.396246	0.100235	0.976535	04:14

In [12]: acc = learn.validate()[1] print(f'Accuracy: {acc:.2%}') Accuracy: 97.74% In []:

epoch train loss valid loss accuracy time 04:14