

```

from fastai.vision.all import *
from fastai.vision.augment import _grid

path = untar_data(URLs.IMAGENETTE_320)

img = Image.open(path.ls()[0].ls()[3].ls()[0])
img = TensorImage(image2tensor(img)[None]/255.)

aug_transforms(max_lighting=.99)[1]

```

Brightness -- {'max\_lighting': 0.99, 'p': 1.0, 'draw': None, 'batch': False}:  
 encodes: (TensorImage,object) -> encodes  
 decodes:

```

show_images(aug_transforms(max_lighting=.99)[1](img.repeat(12,1,1,1)))

```



## RandTransform

Notice that p, can be set to control the probability of a transform being applied.

```

class RandTransform(DisplayedTransform):
    "A transform that before_call its state at each `__call__`"
    do,nm,supports,split_idx = True,None,[],0
    def __init__(self,
        p:float=1., # Probability of applying Transform
        nm:str=None,
        before_call:callable=None, # Optional batchwise preprocessing function
        **kwargs
    ):
        store_attr('p')
        super().__init__(**kwargs)
        self.before_call = ifnone(before_call,self.before_call)

    def before_call(self,
        b,

```

```

        split_idx:int, # Index of the train/valid dataset
    ):
        "This function can be overridden. Set `self.do` based on `self.p`"
        self.do = self.p==1. or random.random() < self.p

    def __call__(self,
        b,
        split_idx:int=None, # Index of the train/valid dataset
        **kwargs
    ):
        self.before_call(b, split_idx=split_idx)
        return super().__call__(b, split_idx=split_idx, **kwargs) if self.do else b

```

## Affine

Fastai has many affine transforms. These include crop, zoom, flip etc. Lets go through some now.

```
xy_grid.shape
```

```
torch.Size([1, 320, 480, 2])
```

```

def show_grid(xy_grid):
    neutral_dim=torch.zeros_like(xy_grid)[...,0,None]
    normal_grid=torch.cat((xy_grid,neutral_dim),dim=3)
    bad_mask = (normal_grid>1).int() + (normal_grid<-1).int()
    bad_mask=-bad_mask.sum(-1)*10
    normal_grid+=bad_mask[...,None]
    show_images(((normal_grid+1)/2).clip(0,1))

```

```

y_coords=torch.linspace(-1,1,img.shape[-2])
x_coords=torch.linspace(-1,1,img.shape[-1])
xy_grid=torch.meshgrid(x_coords, y_coords, indexing='xy')
xy_grid=torch.stack(xy_grid,dim=2)[None]
show_grid(xy_grid)

```



`xy_grid`

```
tensor([[[[-1.0000, -1.0000],
          [-0.9958, -1.0000],
          [-0.9916, -1.0000],
          ...,
          [ 0.9916, -1.0000],
          [ 0.9958, -1.0000],
          [ 1.0000, -1.0000]],

        [[-1.0000, -0.9937],
          [-0.9958, -0.9937],
          [-0.9916, -0.9937],
          ...,
          [ 0.9916, -0.9937],
          [ 0.9958, -0.9937],
          [ 1.0000, -0.9937]],

        [[-1.0000, -0.9875],
          [-0.9958, -0.9875],
          [-0.9916, -0.9875],
          ...,
          [ 0.9916, -0.9875],
          [ 0.9958, -0.9875],
          [ 1.0000, -0.9875]],

        ...,

        [[-1.0000,  0.9875],
          [-0.9958,  0.9875],
          [-0.9916,  0.9875],
```

```

...,
[ 0.9916,  0.9875],
[ 0.9958,  0.9875],
[ 1.0000,  0.9875]],

[[-1.0000,  0.9937],
 [-0.9958,  0.9937],
 [-0.9916,  0.9937],
 ...,
 [ 0.9916,  0.9937],
 [ 0.9958,  0.9937],
 [ 1.0000,  0.9937]],

[[-1.0000,  1.0000],
 [-0.9958,  1.0000],
 [-0.9916,  1.0000],
 ...,
 [ 0.9916,  1.0000],
 [ 0.9958,  1.0000],
 [ 1.0000,  1.0000]]]])

```

```
show_images(F.grid_sample(img,xy_grid))
```



```

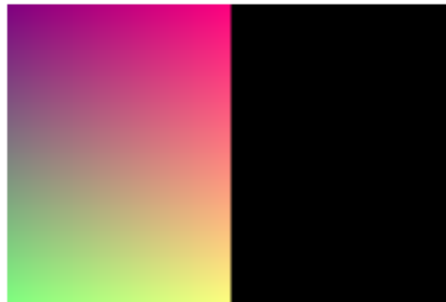
def make_grid(x_coords,y_coords):
    xy_grid=torch.meshgrid(x_coords, y_coords, indexing='xy')
    xy_grid=torch.stack(xy_grid,dim=2)[None]
    return xy_grid

```

```
y_coords=torch.linspace(-1,1,img.shape[-2])
x_coords=torch.linspace(-1,1,img.shape[-1])
make_grid(y_coords)
show_grid(xy_grid)
```

### Slide Left

```
show_grid(make_grid(x_coords+1,y_coords),)
```



```
show_images(F.grid_sample(img,make_grid(-1*x_coords,y_coords)))
```



### Flip

```
show_grid(make_grid(-1*x_coords,y_coords),)
```



```
show_images(F.grid_sample(img,make_grid(2.*x_coords,y_coords)))
```



**Squish/Resize x dim**

```
show_grid(make_grid(2*x_coords,y_coords),)
```



```
show_images(F.grid_sample(img,make_grid(2*x_coords,y_coords)))
```



```
y_coords.shape
```

```
torch.Size([320])
```

```
x_coords.shape
```

```
torch.Size([480])
```

### Why can't we use the technique we have used to implement rotate/warp?

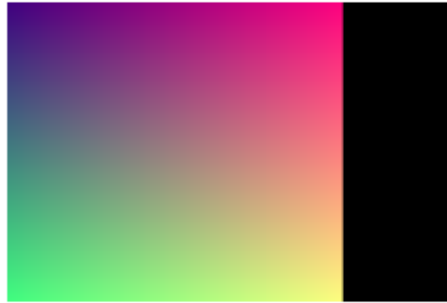
I mostly introduced the previous techniques to make things easy to understand by making x and y independent, but affine transformations can work off of the current x and y values, which takes a bit more code to implement. Lets jump straight into building these like fastai.

### F.affine\_grid

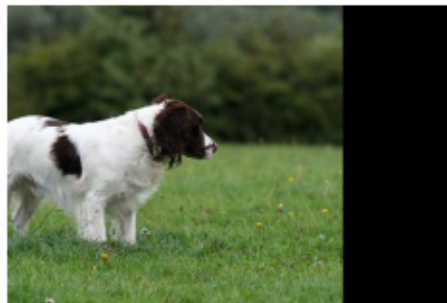
Affine grids work on much smaller grids.

```
translate_grid=torch.tensor([[1.,0,.5],  
                             [0,1,0]])
```

```
coords_grid=F.affine_grid(translate_grid[None], img.shape)  
show_grid(coords_grid)
```



```
show_images(F.grid_sample(img,coords_grid))
```



[https://en.wikipedia.org/wiki/Affine\\_transformation](https://en.wikipedia.org/wiki/Affine_transformation)

```
torch.tensor([[1,0,0],
              [0,1,0]])
```

```
tensor([[1, 0, 0],
        [0, 1, 0]])
```

```
torch.tensor([[1.,0,0],
              [0,1,0]])
```

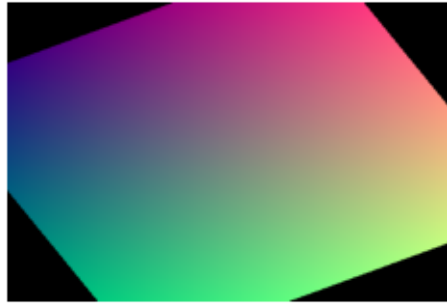
```
tensor([[1., 0., 0.],
        [0., 1., 0.]])
```



```
identity_grid=torch.tensor([[1.,0,0],
                             [0,1,0]])
coords_grid=F.affine_grid(identity_grid[None], img.shape)
show_grid(coords_grid)
show_images(F.grid_sample(img,coords_grid))
```



```
rot_grid=torch.tensor([[math.cos(.5),-math.sin(.5),0],
                        [math.sin(.5),math.cos(.5),0]])
coords_grid=F.affine_grid(rot_grid[None], img.shape)
show_grid(coords_grid)
show_images(F.grid_sample(img,coords_grid))
```

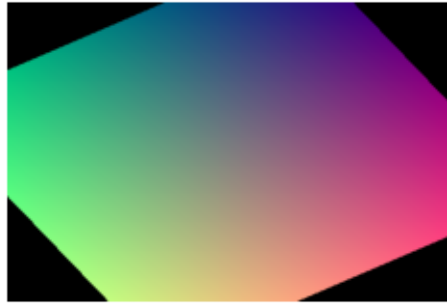


## Rotate

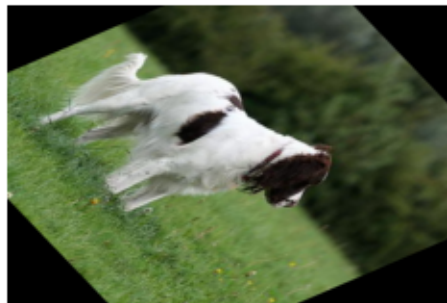
```
math.cos(.5)
```

0.8775825618903728

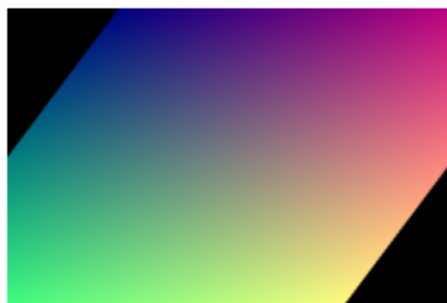
```
rotate_grid=torch.tensor([[math.cos(1),math.sin(1),0],  
                           [-math.sin(1),math.cos(1),0]])  
coords_grid=F.affine_grid(rotate_grid[None], img.shape)  
show_grid(coords_grid)
```



```
show_images(F.grid_sample(img,coords_grid))
```



```
shear_grid=torch.tensor([[1,.5,0],
                          [0,1,0]])
coords_grid=F.affine_grid(shear_grid[None], img.shape)
show_grid(coords_grid)
```

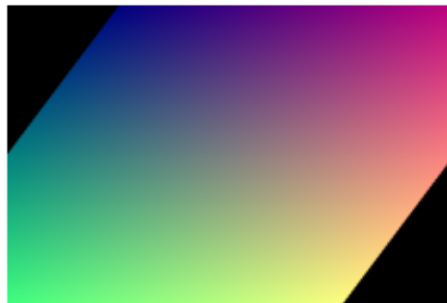


```
show_images(F.grid_sample(img,coords_grid))
```



**How would we warp/skew?**

```
warp_grid=torch.tensor([[1.,.5,0],  
                        [0,1,0]])  
coords_grid=F.affine_grid(warp_grid[None], img.shape)  
show_grid(coords_grid)
```



```
show_images(F.grid_sample(img,coords_grid))
```



## Combining affine augmentations

Lets look at the affine grid identity.

```
affine_grid=torch.tensor([[1.,0,0],
                          [0,1,0]])
coords_grid=F.affine_grid(affine_grid[None], img.shape)
show_grid(coords_grid)
```



Does this affine grid identity look familiar? Can you think of a way to combine affine transforms?

## Implementation

```
def combine_affines(affines):
    id_row=lambda a:torch.cat((a,torch.tensor([.0,0,1])[None]))
    comb_mat=id_row(affines[0])
    for a in affines:
        comb_mat@=id_row(a)
    return comb_mat[:2]
```

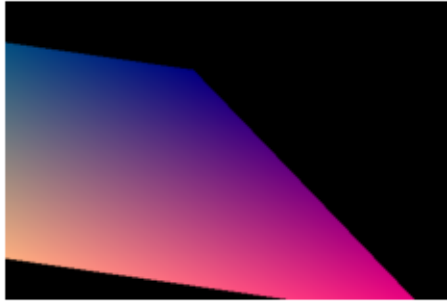
```
wrt_grid=combine_affines([warp_grid,rotate_grid,translate_grid])
```

```
wrt_grid
```

```
tensor([[ -0.3012,  1.3818, -0.3012],
        [-0.8415,  0.5403, -0.8415]])
```

```
coords_grid=F.affine_grid(wrt_grid[None], img.shape)
```

```
show_grid(coords_grid)
```



```
show_images(F.grid_sample(img,coords_grid)),show_images(img)
```

(None, None)



```
_BrightnessLogit??
```

Object `\_BrightnessLogit` not found.

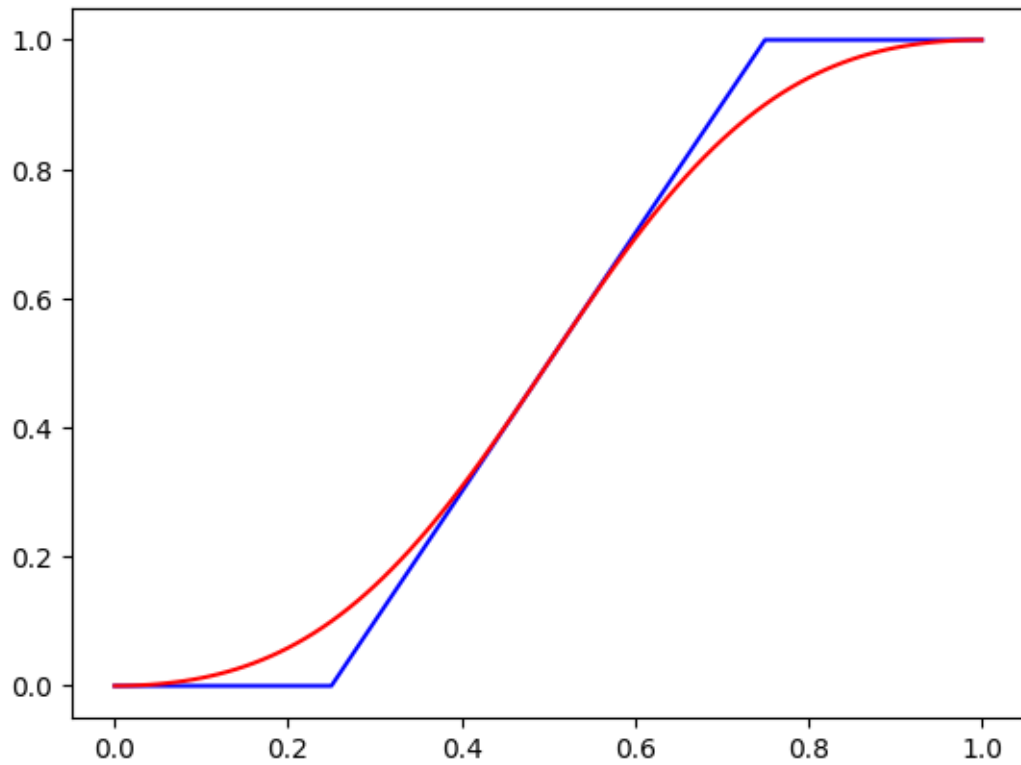
## Lighting

```
show_images((img+.4))
```

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255



```
x=TensorImage(torch.tensor([.01* i for i in range(0,101)]))  
f_lin= lambda x:(2*(x-.5)+.5).clamp(0,1) #blue line  
f_log= lambda x:2*x #red line  
plt.plot(x,f_lin(x),'b',x,torch.sigmoid(f_log(logit(x))),'r');
```



What is special about logit in relationship to sigmoid?

```
show_images(img+.5)
```

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255]





```
show_images(torch.sigmoid(logit(img)+logit(torch.tensor(.85))))
```



```
show_images(img)
```



```
logit??
```

### How to do contrast?

```
show_images(torch.sigmoid(logit(img)*4))
```



## **Next Section**

Open other notebook