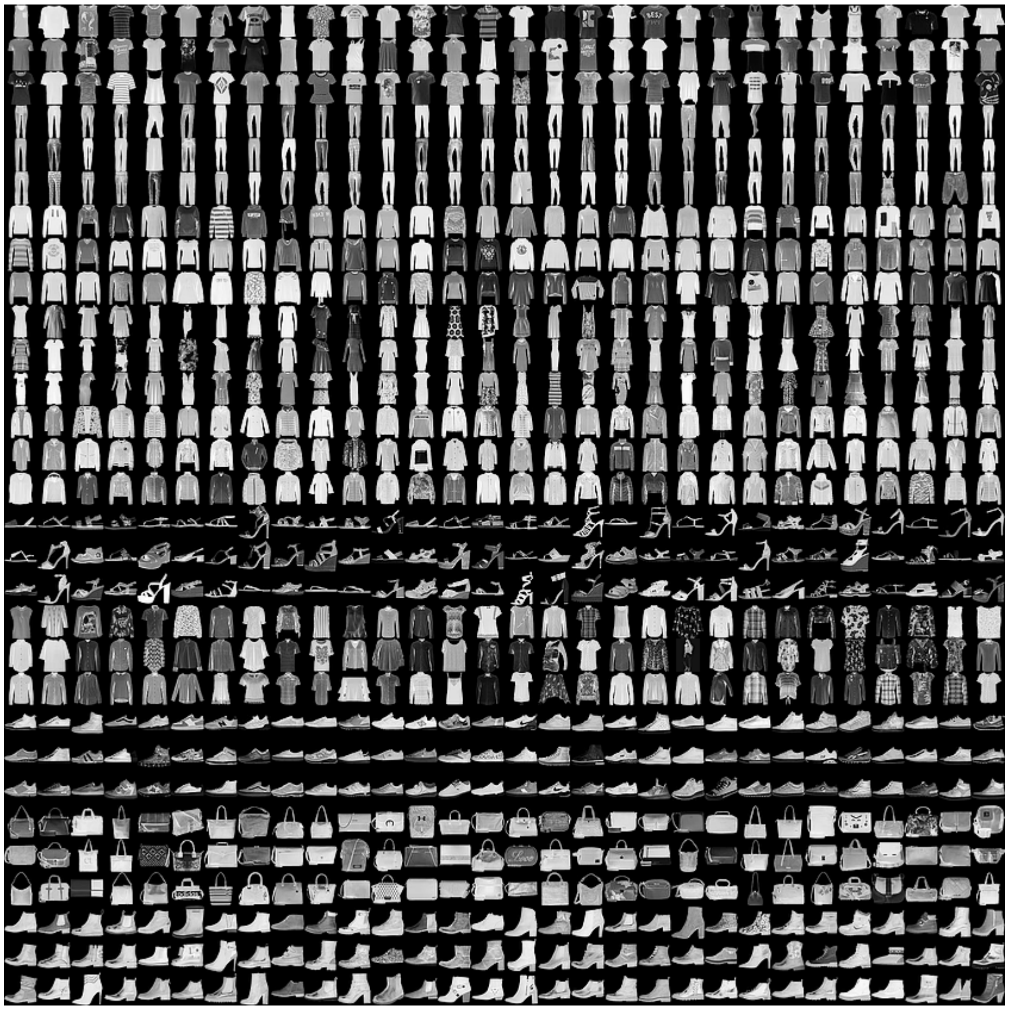
Classifying Fashion-MNIST

Now it's your turn to build and train a neural network. You'll be using the Fashion-MNIST dataset (https://github.com/zalandoresearch/fashion-mnist), a drop-in replacement for the MNIST dataset. MNIST is actually quite trivial with neural networks where you can easily achieve better than 97% accuracy. Fashion-MNIST is a set of 28x28 greyscale images of clothes. It's more complex than MNIST, so it's a better representation of the actual performance of your network, and a better representation of datasets you'll use in the real world.



In this notebook, you'll build your own neural network. For the most part, you could just copy and paste the code from Part 3, but you wouldn't be learning. It's important for you to write the code yourself and get it to work. Feel free to consult the previous notebooks though as you work through this.

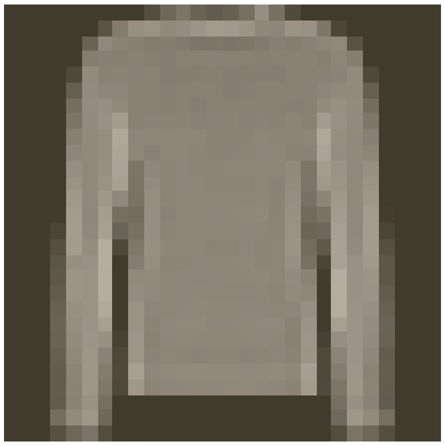
First off, let's load the dataset through torchvision.

In [1]:
```
import torch
from torchvision import datasets, transforms
import helper

# Define a transform to normalize the data
transform = transforms.Compose([transforms.ToTensor(),
                                transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.
5, 0.5))])
# Download and load the training data
trainset = datasets.FashionMNIST('~/.pytorch/F_MNIST_data/', download=True, tr
ain=True, transform=transform)
trainloader = torch.utils.data.DataLoader(trainset, batch_size=64, shuffle=Tru
e)

# Download and load the test data
testset = datasets.FashionMNIST('~/.pytorch/F_MNIST_data/', download=True, tra
in=False, transform=transform)
testloader = torch.utils.data.DataLoader(testset, batch_size=64, shuffle=True)
```

```
Downloading http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/train-
images-idx3-ubyte.gz
Downloading http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/train-
labels-idx1-ubyte.gz
Downloading http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/t10k-i
mages-idx3-ubyte.gz
Downloading http://fashion-mnist.s3-website.eu-central-1.amazonaws.com/t10k-l
abels-idx1-ubyte.gz
Processing...
Done!
```

Here we can see one of the images.

In [2]:
```
image, label = next(iter(trainloader))
helper.imshow(image[0,:]);
```



Building the network

Here you should define your network. As with MNIST, each image is 28x28 which is a total of 784 pixels, and there are 10 classes. You should include at least one hidden layer. We suggest you use ReLU activations for the layers and to return the logits or log-softmax from the forward pass. It's up to you how many layers you add and the size of those layers.

In [3]:
```
# TODO: Define your network architecture here
from torch import nn
from torch import optim

model = nn.Sequential(nn.Linear(784, 200),
                      nn.ReLU(),
                      nn.Linear(200, 80),
                      nn.ReLU(),
                      nn.Linear(80, 10),
                      nn.LogSoftmax(dim=1))
```

Train the network

Now you should create your network and train it. First you'll want to define the criterion (http://pytorch.org/docs/master/nn.html#loss-functions) (something like `nn.CrossEntropyLoss`) and the optimizer (http://pytorch.org/docs/master/optim.html) (typically `optim.SGD` or `optim.Adam`).

Then write the training code. Remember the training pass is a fairly straightforward process:

- Make a forward pass through the network to get the logits
- Use the logits to calculate the loss
- Perform a backward pass through the network with `loss.backward()` to calculate the gradients
- Take a step with the optimizer to update the weights

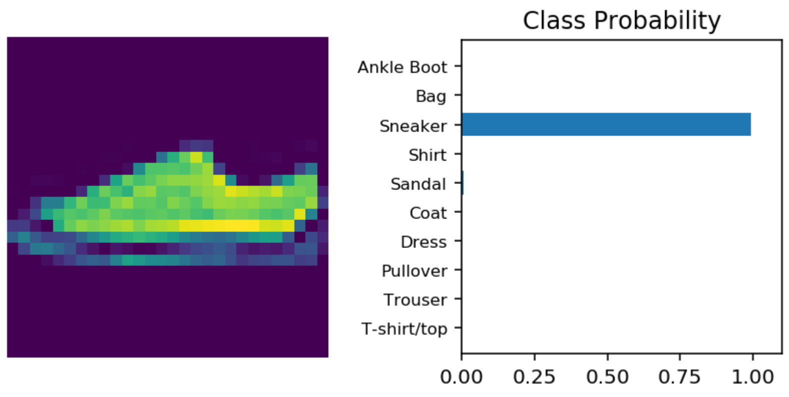
By adjusting the hyperparameters (hidden units, learning rate, etc), you should be able to get the training loss below 0.4.

In [4]:
```
# TODO: Create the network, define the criterion and optimizer
criterion = nn.NLLLoss()
optimizer = optim.SGD(model.parameters(), lr = 0.01)
```

```
In [5]: # TODO: Train the network here

        epochs = 10

        for _ in range(epochs):
            epoch_loss = 0

            for images, labels in trainloader:
                images = images.view(images.shape[0], -1);

                optimizer.zero_grad()
                outputs = model.forward(images)
                loss = criterion(outputs, labels)
                loss.backward()

                optimizer.step()

                epoch_loss += loss.item()
            else:
                print(f'Traing loss: {epoch_loss / len(trainloader)}')
```

```
Traing loss: 1.0013768023837095
Traing loss: 0.5502443253866899
Traing loss: 0.4818135339981203
Traing loss: 0.44881746984684645
Traing loss: 0.4264577081653355
Traing loss: 0.40971493292083616
Traing loss: 0.39600194344070677
Traing loss: 0.38417961499266534
Traing loss: 0.37383431472630896
Traing loss: 0.36506272188381855
```

In [6]:
```
%matplotlib inline
%config InlineBackend.figure_format = 'retina'

import helper
import torch.nn.functional as F

# Test out your network!

dataiter = iter(testloader)
images, labels = dataiter.next()
img = images[0]
# Convert 2D image to 1D vector
img = img.resize_(1, 784)

# TODO: Calculate the class probabilities (softmax) for img
with torch.no_grad():
    ps = F.softmax(model.forward(img), dim = 1)

# Plot the image and probabilities
helper.view_classify(img.resize_(1, 28, 28), ps, version='Fashion')
```


Class Probability
Ankle Boot
Bag
Sneaker
Shirt
Sandal
Coat
Dress
Pullover
Trouser
T-shirt/top