Slice Sampling Unimodal

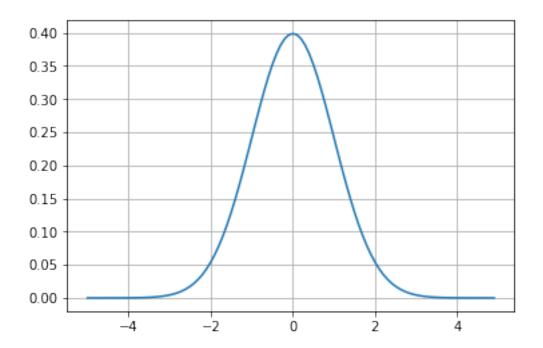
March 21, 2019

1 Unimodal Slice Sampling Example

1.0.1 Import Libraries

```
In [698]: import numpy as np
        import math
        from scipy.stats import norm as norm
    import matplotlib.pyplot as plt
        %matplotlib inline
```

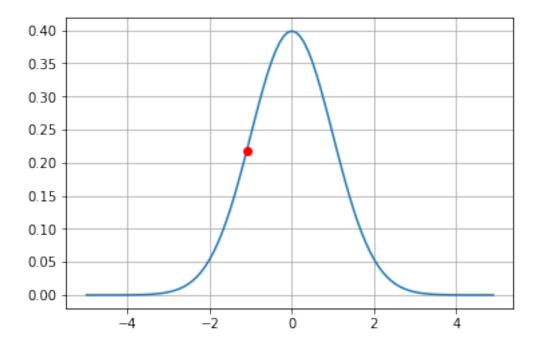
1.0.2 Generate Unimodal Gaussian



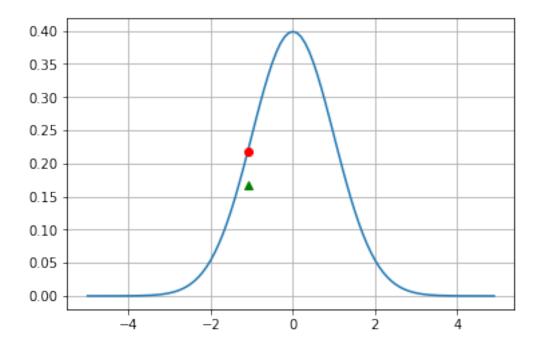
1.0.3 Define function for initial sample for x0

1.0.4 Define function for calculating f(x0)

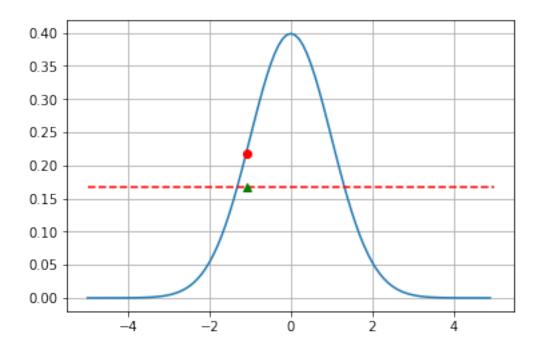
```
plt.plot([x0], [f_x0], 'ro')
plt.grid(True)
plt.show()
```



1.0.5 Define function for sampling in the interval $(0, f(x_0))$, calculate y



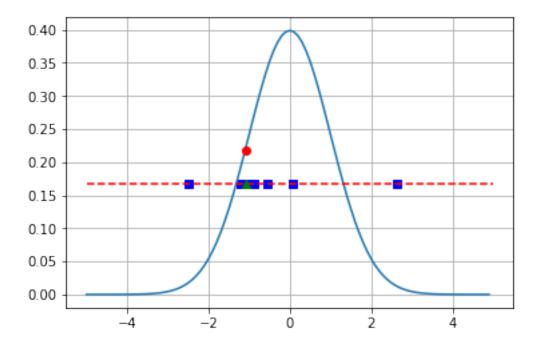
1.0.6 Define horizontal slice using y



1.0.7 Define function for estimating the sampling interval using doubling update

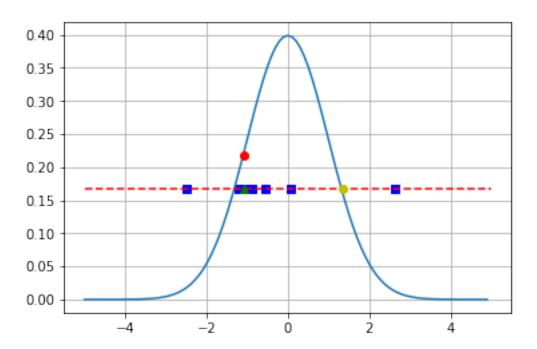
```
In [705]: def doubling_update(x0, f_x0, y, w=0.01, p=100):
              u = np.random.uniform()
              1 = x0 - w*u
              r = 1 + w
              k = p
              left = []
              right = []
              while k > 0 and (y < f_x(1) \text{ or } y < f_x(r)):
                  v = np.random.uniform()
                  if v < .5:
                      1 = 1 - (r-1)
                      left.append(1)
                  else:
                      r = r + (r-1)
                      right.append(r)
              patches = np.concatenate((left,right), axis=None)
              return 1, r, patches
          double = doubling_update(x0, f_x0, y)
          ascisse = np.array([y for i in range(len(double[2]))])
```

```
plt.plot(double[2],np.array([y for i in range(len(double[2]))]) , 'bs')
plt.plot(X, Y)
plt.plot(line[0], line[1], 'r--')
plt.plot([x0], [f_x0], 'ro')
plt.plot([x0], [y], 'g^')
plt.grid(True)
plt.show()
```



1.0.8 Define updating rule for sampling x_n from new interval

```
plt.grid(True)
plt.show()
```



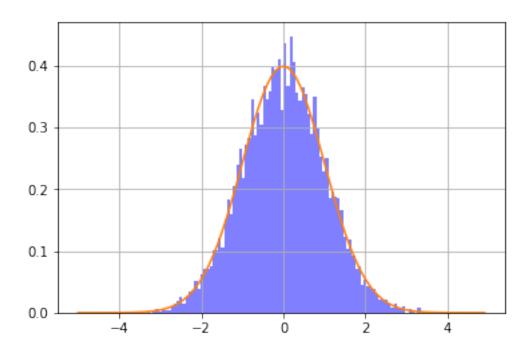
1.0.9 Sample and plot results

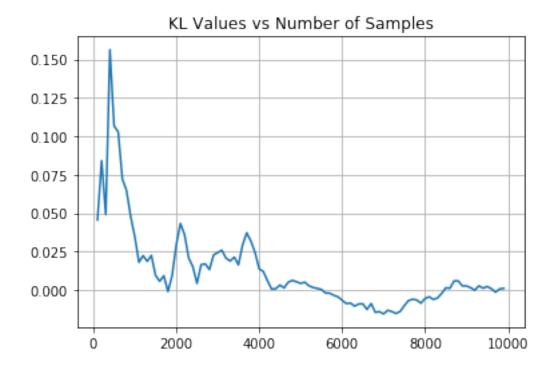
```
In [721]: samples_double = []
          iterations_d = []
          kl_values_d = []
          i = 0
          1 = double[0]
          r = double[1]
          while i <10000:
              new_x = new_x_double(y, 1, r)
              new_fx = f_x(new_x)
              new_sampled_y = sample_y(new_x, new_fx)
              new_double = doubling_update(new_x, new_fx, new_sampled_y)
              samples_double.append(round(new_x,2))
              y = new_sampled_y
              1 = new_double[0]
              r = new_double[1]
              if i%100==0:
                  (mu_d, sigma_d) = norm.fit(samples_double)
                  kl\_values\_d.append(np.log(sigma\_d/1)+((1+(0-mu\_d)**2)/2*sigma\_d**2)-.5)
                  iterations_d.append(i)
              i = i+1
```

```
num_bins = 100
n_d, bins_d, patches_d = plt.hist(samples_double, num_bins, facecolor='blue', alpha=
plt.plot(X, Y)
plt.grid(True)
plt.show()

plt.title(r'KL Values vs Number of Samples')
plt.plot(iterations_d, kl_values_d)
plt.grid(True)
plt.show()
```

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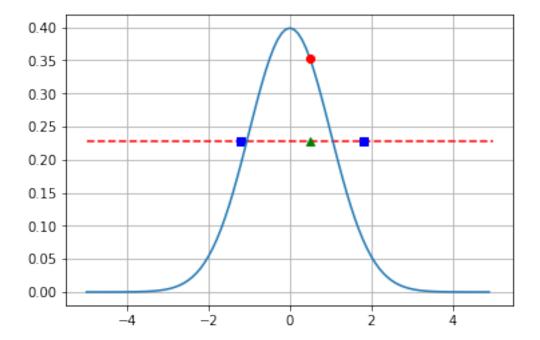




1.0.10 Define Stepout Update function

```
In [645]: def stepout_update(x0, f_x0, y, w=1, m=40):
              u = np.random.uniform()
              1 = x0 - w*u
              r = 1 + w
              v = np.random.uniform()
              j = math.floor(m*v)
              k = (m-1) - j
              patches = []
              while j>0 and y<f_x(1):
                  1 -= w
                  j -= 1
                  patches.append(1)
              while k>0 and y<f_x(r):
                  r += w
                  k = 1
                  patches.append(r)
              return 1, r, patches
          stepout = stepout_update(x0, f_x0, y)
```

```
line = slice_y(y)
plt.plot(line[0], line[1], 'r--')
plt.plot(stepout[2], np.array([y for i in range(len(stepout[2]))]), 'bs')
plt.plot(X, Y)
plt.plot([x0], [f_x0], 'ro')
plt.plot([x0], [y], 'g^')
plt.grid(True)
plt.show()
```



1.0.11 Define updating rule for sampling x_n from new interval

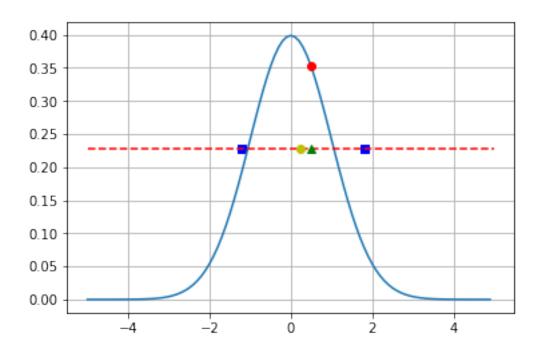
```
In [646]: def update_x_stepout(y, 1, r):
    new_x = sample_x(1, r)
    while y > f_x(new_x):
        new_x = sample_x(1, r)
    return new_x

new_x_stepout = update_x_stepout(y, stepout[0], stepout[1])
print(new_x_stepout)

plt.plot(stepout[2], np.array([y for i in range(len(stepout[2]))]), 'bs')
plt.plot(X, Y)
plt.plot(line[0], line[1], 'r--')
plt.plot([x0], [f_x0], 'ro')
plt.plot([x0], [y], 'g^')
plt.plot([new_x_stepout], [y], 'yo')
```

```
plt.grid(True)
plt.show()
```

0.23614233002285268



1.0.12 Sample and plot results and calculate KL Divergence

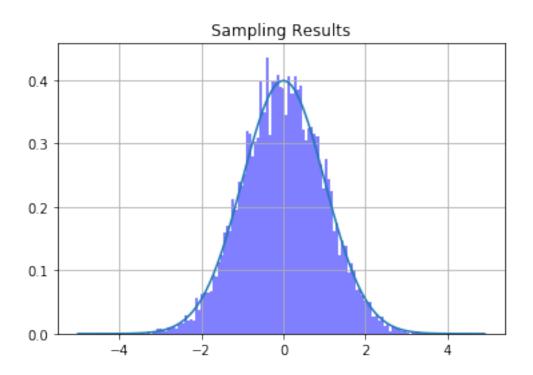
```
In [694]: samples_stepout = []
          iterations_s = []
          kl_values_s = []
          i = 0
          1 = stepout[0]
          r = stepout[1]
          while i <10000:
              new_x = update_x_stepout(y, 1, r)
              new_fx = f_x(new_x)
              new_sampled_y = sample_y(new_x, new_fx)
              new_stepout = stepout_update(new_x, new_fx, new_sampled_y)
              samples_stepout.append(round(new_x,2))
              y = new_sampled_y
              1 = new_stepout[0]
              r = new_stepout[1]
              if i%100==0:
                  (mu_s, sigma_s) = norm.fit(samples_stepout)
```

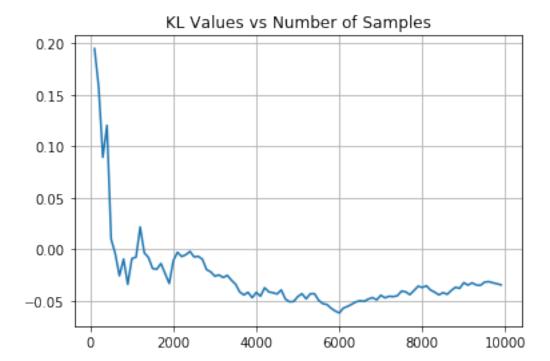
```
kl_values_s.append(np.log(sigma_s/1)+((1+(0-mu_s)**2)/2*sigma_s**2)-.5)
    iterations_s.append(i)
    i = i+1

plt.title(r'Sampling Results')
plt.plot(X, Y)
num_bins = 100
n_s, bins_s, patches_s = plt.hist(samples_stepout, num_bins, facecolor='blue', alpha:plt.grid(True)
plt.show()

plt.title(r'KL Values vs Number of Samples')
plt.plot(iterations_s, kl_values_s)
plt.grid(True)
plt.show()
```

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1.0.13 Fit approximate Gaussians to Sample Distributions

