```
In [28]:
 | import numpy as np
    import matplotlib.pyplot as plt
    import os.path
    import pandas as pd
    import scipy
    # Load data
    npz = np.load('heart_disease_data.npz')
    df = pd.DataFrame.from_dict({item: npz[item] for item in npz.files}, orient='
    train = df.iloc[0:50,:]
    test = df.iloc[50:,:]
    def ind_x_eq_val(x, val):
        return np.where(x==val)[0]
    def count x eq val(x, val):
        return len(ind_x_eq_val(x, val))/float(len(x))
    def gaussian(x, mu, sig):
        return np.exp(-np.power(x - mu, 2.) / (2 * np.power(sig, 2.))) / sig / np
```

QUESTION (b)

Non-parametric model Compute empirical pmf, derive the conditional pmf, and estimate the MAP decision by the mode of posterior distribution $p_{\tilde{h}|\tilde{s},\tilde{c}}$. (The MAP estimates should be $\tilde{h}=0 \text{ or } 1$)

```
In [41]:
 \parallel # Estimate the pmf of H, i.e. P(H=0) and P(H=1)
    P H = np.zeros(2)
    for i in range(2):
        P_H[i] = count_x_eq_val(df['heart_disease'],i)
        if i == 0:
             P H0 = count x eq val(df['heart disease'],i)
        else:
            P H1 = count x eq val(df['heart disease'],i)
    # Estimate the conditional pmf of S given H, i.e. P(S|H=0) and P(S|H=1)
    P S H0 = np.zeros(2)
    P_S_H1 = np.zeros(2)
    for ind S in range(2):
        P S H0[ind S] = len(df[(df['heart disease'] == 0) & (df['sex']==ind S)])
        P_S_H1[ind_S] = len(df[(df['heart_disease'] == 1) & (df['sex']==ind_S)])
    # Estimate the conditional pmf of C given H, i.e. P(C|H=0) and P(C|H=1)
    P C H0 = np.zeros(4)
    P_C_H1 = np.zeros(4)
    for ind C in range(4):
        P_C_H0[ind_C] = len(df[(df['heart_disease'] == 0) & (df['chest_pain']==in
        P C H1[ind C] = len(df[(df['heart disease'] == 1) & (df['chest pain']==in
    h0 list = []
    for s in range(2):
        for c in range(4):
            top = P_H[0]*P_S_H0[s]*P_C_H0[c]
            h0_list.append(top)
    h1_list = []
    for s in range(2):
        for c in range(4):
            top = P_H[1]*P_S_H1[s]*P_C_H1[c]
            h1_list.append(top)
    map estimate S C = []
    for i in range(len(h0 list)):
        if h1 list[i] < h0 list[i]:</pre>
            map_estimate_S_C.append(0)
        else:
            map estimate S C.append(1)
    errors = 0
    index = 0
    df = df.iloc[0:50,:]
    for s in range(2):
        for c in range(4):
            df_of_interest = df[(df['sex_test']==s) & (df['chest_pain_test']==c)]
            for i in df of interest['heart disease test']:
                 if i != map estimate S C[index]:
                     errors += 1
            index += 1
    error rate S C = errors / len(df)
```

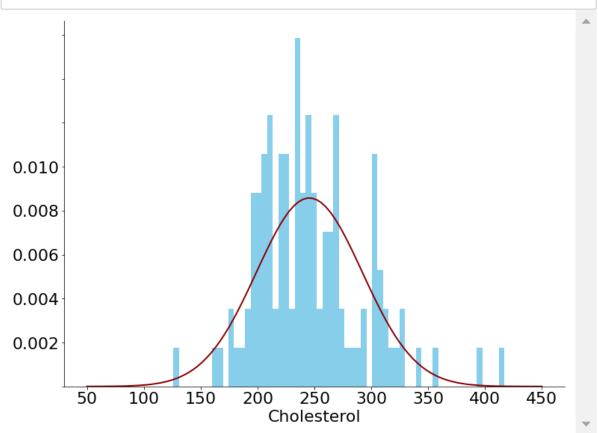
```
print("Probability of error " + str(error_rate_S_C))#
```

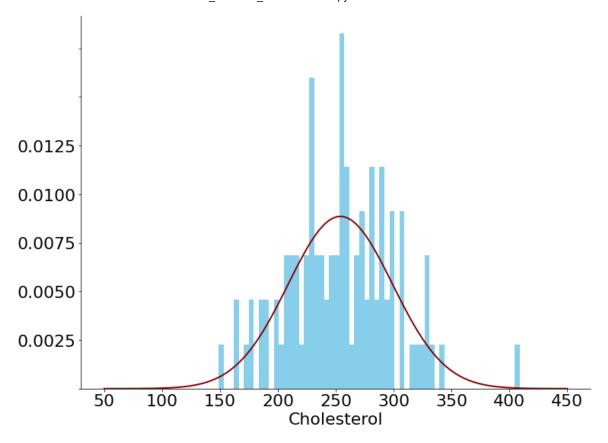
Probability of error 0.18

QUESTION (d)

Maximum likelihood estimates Find the parameters of two normal distributions ($\tilde{x}|\tilde{h}=1$ and $\tilde{x}|\tilde{h}=0$) that maximize the likelihood functions.

```
In [42]:
## Estimate MLE of X given H
   no heart cond = (df[df['heart disease']==0])
   heart_cond = (df[df['heart_disease']==1])
   mean_X_H = np.zeros(2)
   std X H = np.zeros(2)
   mean_X_H[0] = np.mean(np.array(no_heart_cond['cholesterol']))
   std_X_H[0] = np.std(np.array(no_heart_cond['cholesterol']))
   mean_X_H[1]= np.mean(np.array(heart_cond['cholesterol']))
   std_X_H[1] = np.std(np.array(heart_cond['cholesterol']))
   n_plot = 100
   for i in range(2):
       plt.figure(figsize=(12, 9))
       ax = plt.subplot(111)
       ax.spines["top"].set_visible(False)
       ax.spines["right"].set_visible(False)
       ax.get_xaxis().tick_bottom()
       ax.get_yaxis().tick_left()
       yticks = ax.yaxis.get_major_ticks()
       yticks[0].label1.set_visible(False)
       plt.xticks(fontsize=22)
       plt.yticks(fontsize=22)
       plt.xlabel("Cholesterol", fontsize=22)
       plt.hist(df.loc[df['heart disease']==i]['cholesterol'],
               bins=60, stacked=True, edgecolor = "none", color="skyblue", densit
       plt.plot(np.linspace(50, 450, n_plot),gaussian(np.linspace(50, 450, n_plot))
                       mean_X_H[i], std_X_H[i]), color="darkred", lw=2)
```





QUESTION(e)

MAP decision compute posterior $p_{\tilde{h}|\tilde{s},\tilde{c},\tilde{\chi}}$ and derive MAP

In [43]: ▶ from scipy import stats

```
In [44]:
 # Calculate the MAP estimate
    MAP_list = []
    for s in range(2):
        for c in range(4):
            for chol in set(df['cholesterol_test']):
                pdf_value_H0 = scipy.stats.norm.pdf(chol, mean_X_H[0], std_X_H[0]
                pdf value H1 = scipy.stats.norm.pdf(chol, mean X H[1], std X H[1]
                var1 = P H1*P S H1[s]*P C H1[c]*pdf value H1
                var2 = P_H0*P_S_H0[s]*P_C_H0[c]*pdf_value_H0
                if var1 < var2: MAP_list.append(0)</pre>
                else: MAP list.append(1)
    MAP estimate S C X = max(set(MAP list), key=MAP list.count)
    # Calculate the error rate
    list index = 0
    error_count = 0
    for s in range(2):
        for c in range(4):
            for chol in set(df['cholesterol test']):
                sub_df = train[(train['sex_test'] == s) & (train['chest_pain_test
                for idx2 in sub df['heart disease test']:
                    if idx2 != MAP_list[list_index]:
                        error count += 1
                list index += 1
    error_rate_S_C_X = error_count / len(train)
    print("Probability of error using cholesterol " + str(error_rate_S_C_X))
```

Probability of error using cholesterol 0.14

In []:	K	
In []:	M	
In []:	H	