

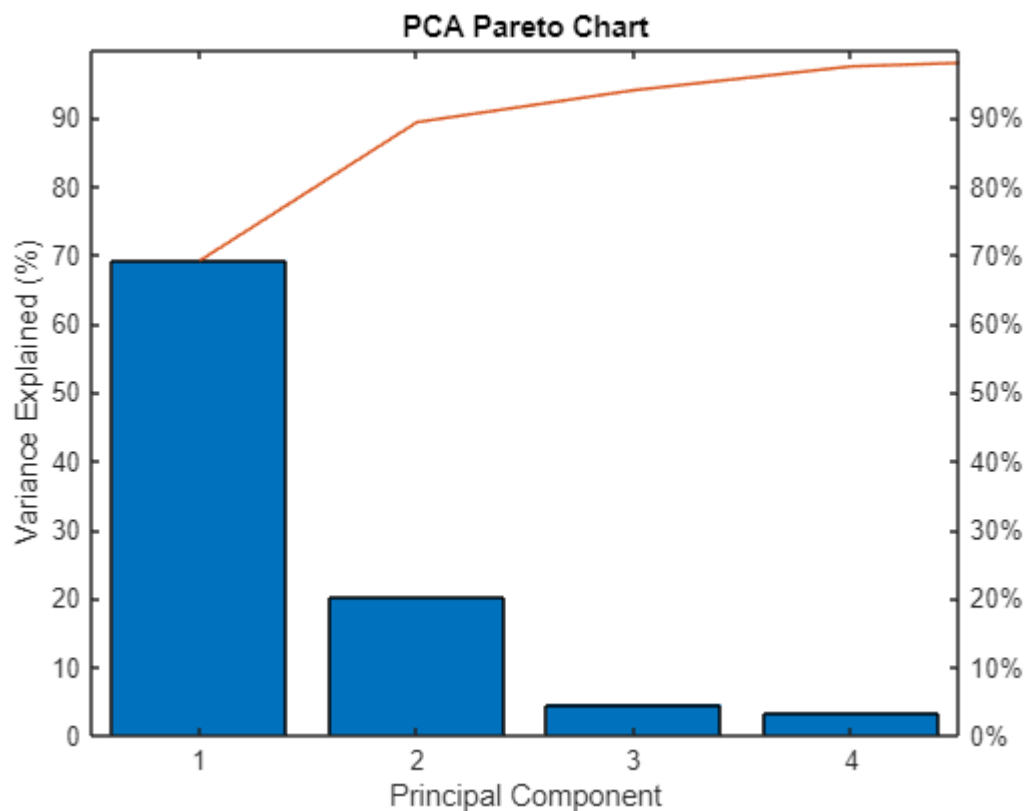
```
% Performing PCA on Training data  
[coeff, scoreTrain, ~, ~, explained, mu] = pca(X_train);
```

Warning: Columns of X are linearly dependent to within machine precision.
Using only the first 28 components to compute TSQUARED.

```
% Select number of components to retain 95% variance  
numPCs = find(cumsum(explained) >= 95, 1)
```

```
numPCs =  
4
```

```
%Plotting Pareto Chart  
figure;  
pareto(explained);  
xlabel('Principal Component');  
ylabel('Variance Explained (%)');  
title('PCA Pareto Chart');
```

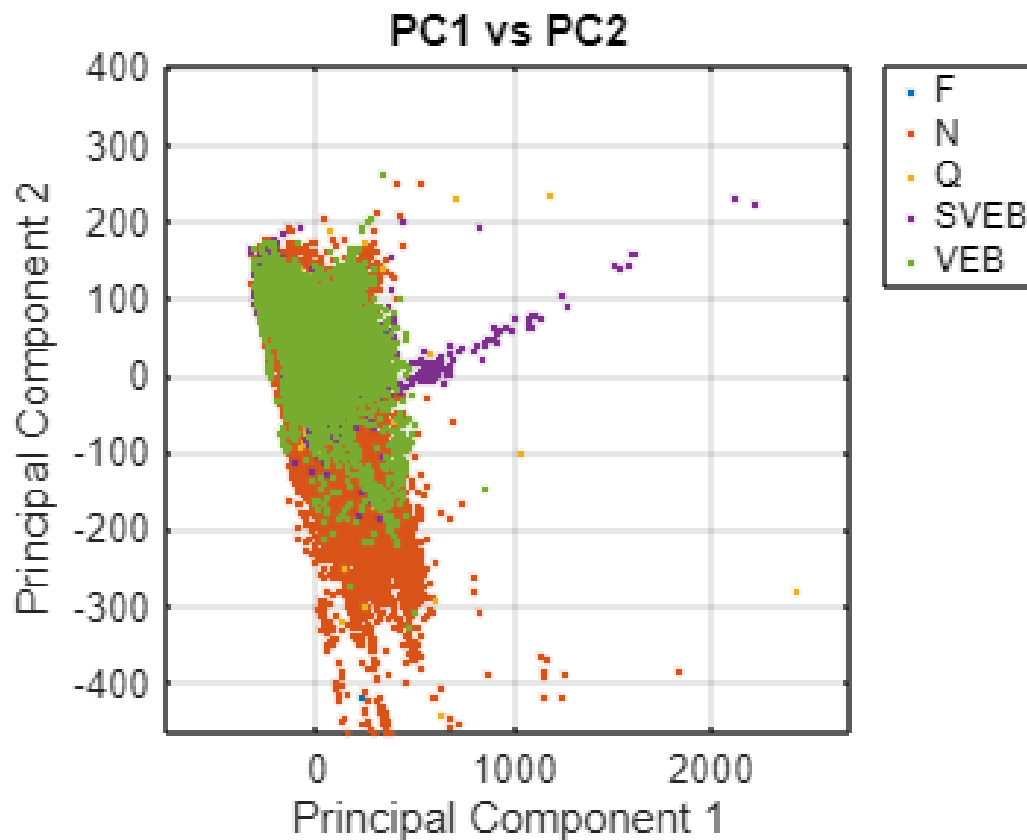


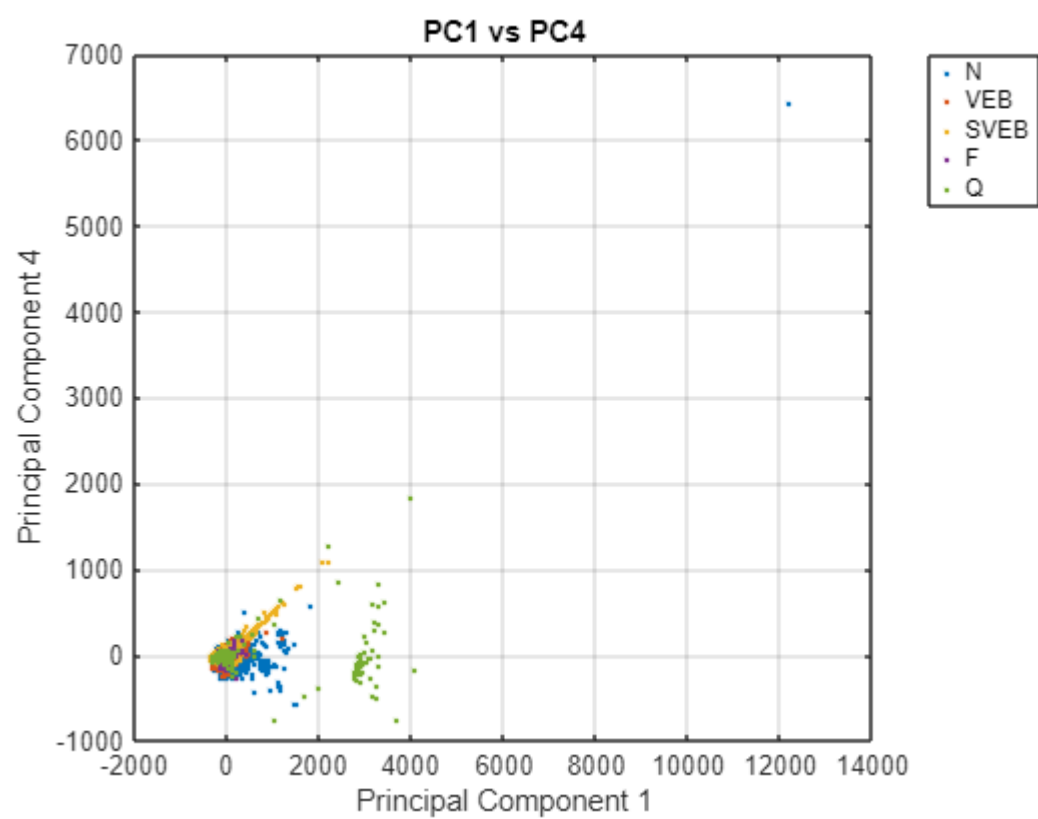
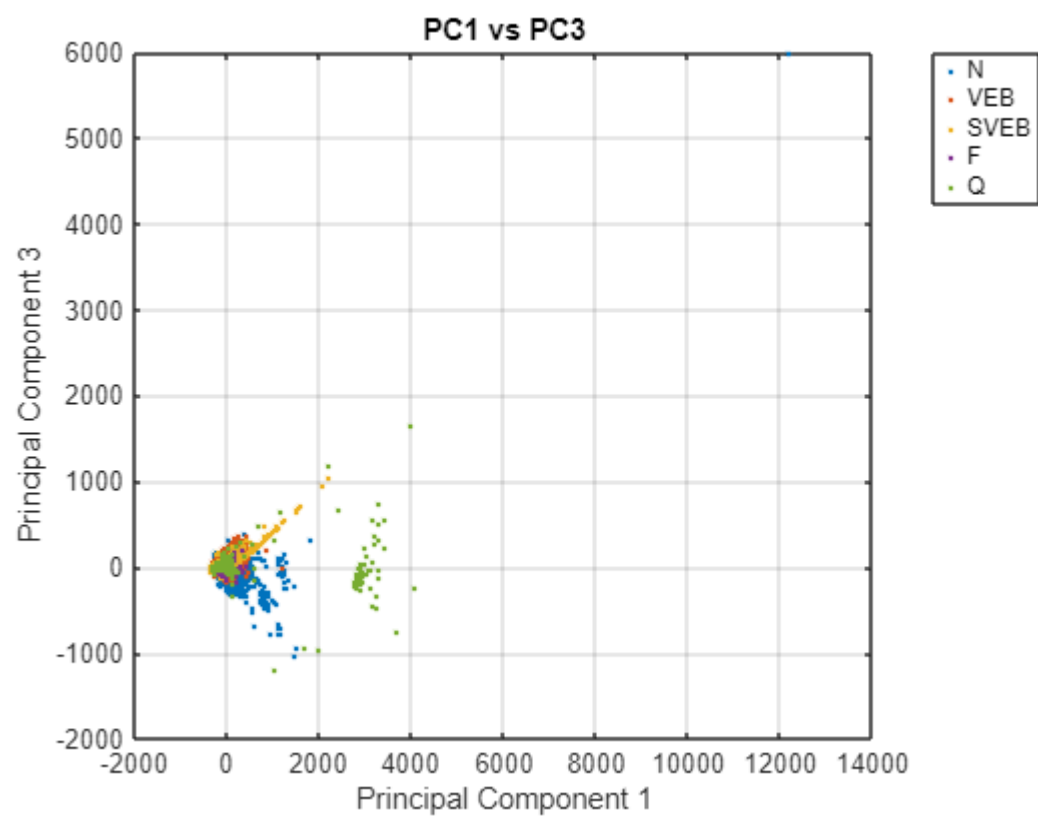
```
% Plot all combinations of the first few principal components (e.g., 4)  
numToPlot = 4; % Adjust this number based on how many PCs you want to explore
```

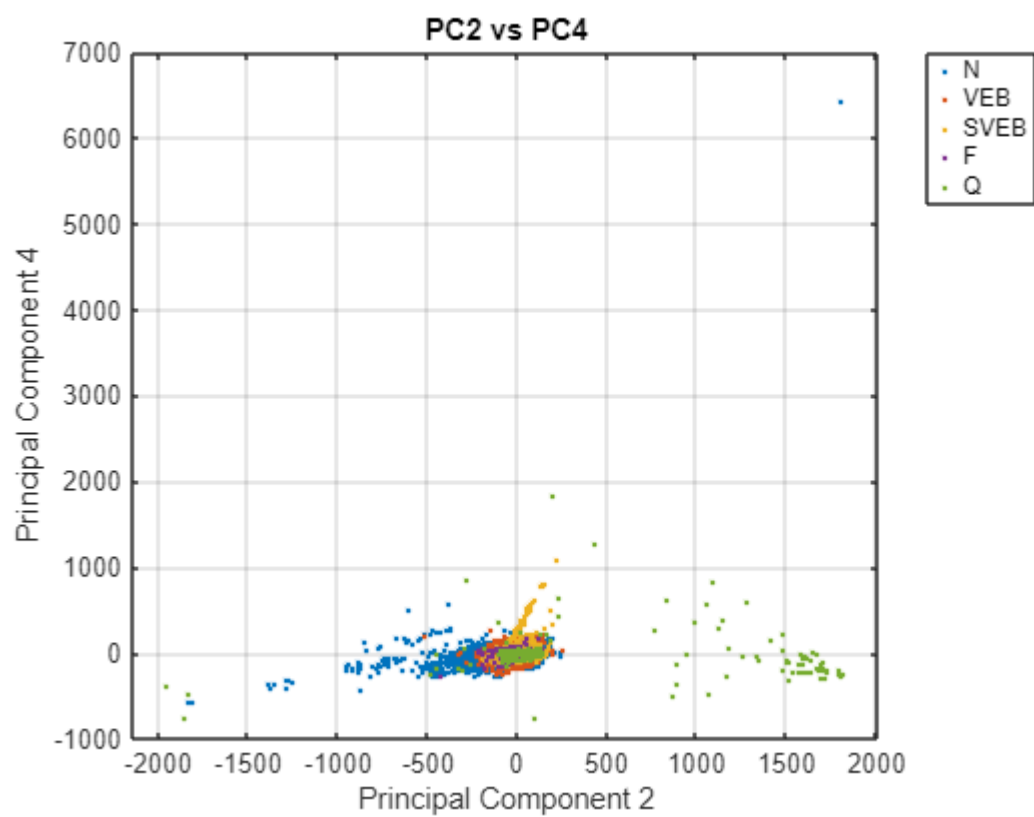
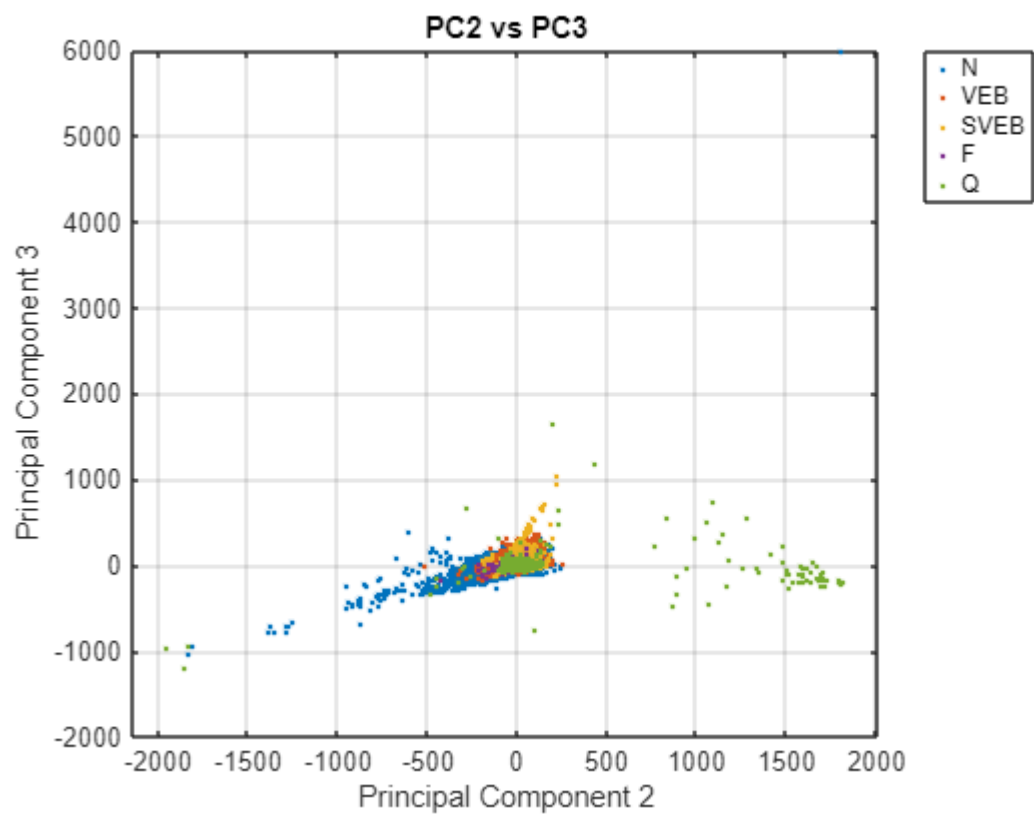
```

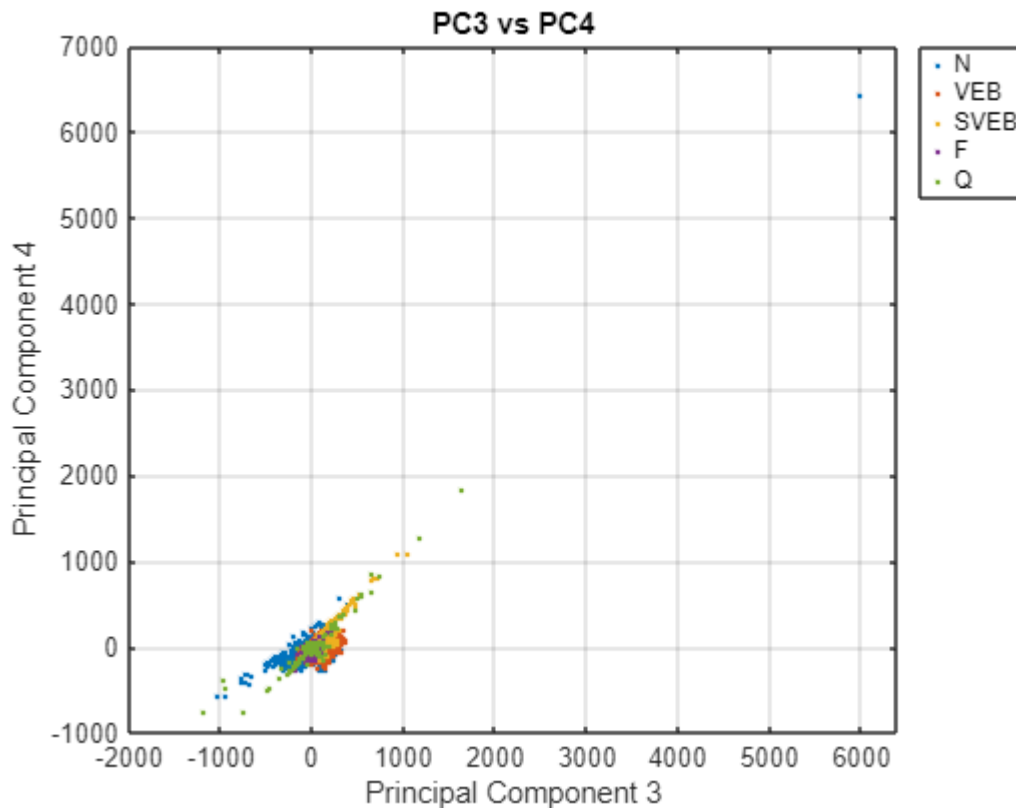
for i = 1:numToPlot
    for j = i+1:numToPlot
        figure;
        gscatter(scoreTrain(:,i), scoreTrain(:,j), Y_train);
        xlabel(['Principal Component ' num2str(i)]);
        ylabel(['Principal Component ' num2str(j)]);
        title(['PC' num2str(i) ' vs PC' num2str(j)]);
        legend('Location','bestoutside');
        grid on;
    end
end

```









```
% Convert Y_train to string if it's not already
Y_train = string(Y_train);

% Unique class labels as a string array
classes = unique(Y_train);

% Define color map
colors = lines(numel(classes));

% 3D PCA plot
figure;
hold on;
for i = 1:numel(classes)
    idx = Y_train == classes(i); % use string comparison
    scatter3(scoreTrain(idx,1), scoreTrain(idx,2), scoreTrain(idx,3), ...
            36, colors(i,:), 'filled', 'DisplayName', classes(i));
end
hold off;
xlabel('Principal Component 1');
ylabel('Principal Component 2');
zlabel('Principal Component 3');
title('3D PCA Projection');
legend('Location', 'bestoutside');
grid on;
view(135, 30);
```

