test

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```
# Load necessary libraries
library(mclust)
## Warning: package 'mclust' was built under R version 4.3.2
## Package 'mclust' version 6.1
## Type 'citation("mclust")' for citing this R package in publications.
library(tidyverse)
## Warning: package 'tidyr' was built under R version 4.3.2
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr 1.1.4 v readr
                                   2.1.5
## v forcats 1.0.0 v stringr 1.5.1
## v ggplot2 3.4.4 v tibble
                                  3.2.1
## v lubridate 1.9.3
                        v tidyr
                                   1.3.1
## v purrr
              1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
                    masks mclust::map()
## x purrr::map()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
# Read the data
df <- read.csv("./data/diabetes.csv") |>
 janitor::clean_names() |>
 mutate(across(c(glucose, blood_pressure, skin_thickness, insulin, bmi), ~na_if(.x, 0)))
# Handle missing values
df[is.na(df)] <- 0</pre>
# Extract features
X <- df[, c("pregnancies", "glucose", "blood_pressure", "skin_thickness", "insulin", "bmi",
           "diabetes_pedigree_function", "age")]
# Define the EM algorithm function
EM_algorithm <- function(X, num_clusters, max_iter = 100, tol = 1e-6) {</pre>
```

```
# Initialize parameters
  model <- Mclust(X, G = num_clusters)</pre>
  for (iter in 1:max iter) {
    # E-step
    responsibilities <- matrix(0, nrow = nrow(X), ncol = num_clusters)</pre>
    for (k in 1:num_clusters) {
      responsibilities[, k] <- model$z[, k] * model$parameters$pro[k]
    }
    responsibilities <- responsibilities / rowSums(responsibilities)
    # M-step
    model <- Mclust(X, G = num_clusters, z = responsibilities)</pre>
    # Calculate log-likelihood and check for convergence
    log_likelihood <- sum(log(apply(model$z * model$parameters$pro, 1, sum)))</pre>
    if (iter > 1 && abs(log_likelihood - prev_log_likelihood) < tol) {</pre>
      break
    }
    prev_log_likelihood <- log_likelihood</pre>
 return(model)
# Call the EM algorithm
num_clusters <- 3  # Set the number of clusters</pre>
gmm_model <- EM_algorithm(X, num_clusters)</pre>
# Get the cluster assignments for each sample
clusters <- predict(gmm_model)</pre>
# View the feature patterns of the clusters
cluster_means <- t(apply(gmm_model$parameters$mean, 2, function(x) round(x, 2)))</pre>
print(cluster_means)
        pregnancies glucose blood_pressure skin_thickness insulin
## [1,]
            4.25 133.59
                                      62.55
                                                      25.76 137.72 32.63
                                      69.86
## [2,]
               3.11 112.17
                                                      28.18 88.94 31.97
## [3,]
               4.86 124.39
                                      74.63
                                                       0.00
                                                                0.00 31.35
        diabetes_pedigree_function
## [1,]
                               0.68 38.07
## [2,]
                               0.40 28.05
## [3,]
                               0.40 38.23
# View the mixing weights of the clusters
mixing_weights <- round(gmm_model$parameters$pro, 2)</pre>
print(mixing_weights)
```

[1] 0.27 0.49 0.25

```
# Further analysis can be performed based on the cluster assignments,
# for example, analyzing the relationship between cluster assignments and diabetes status
```

```
library(ggplot2)

# Ensure the cluster and diabetes_status variables are factors

df$cluster <- factor(clusters$classification)

# Calculate the proportion of outcome (mean) within each cluster

cluster_outcome_mean <- aggregate(outcome ~ cluster, data = df, FUN = mean)

# Plot the distribution of diabetes status within each cluster

ggplot(df, aes(x = cluster, y = outcome, fill = outcome)) +

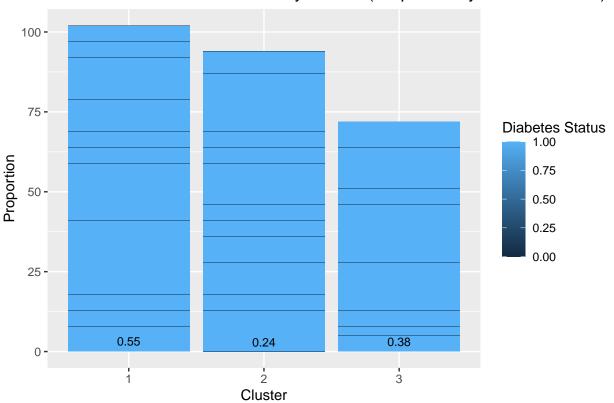
geom_bar(stat = "identity") +

geom_text(data = cluster_outcome_mean, aes(label = sprintf("%.2f", outcome)), vjust = -0.5, color = "

labs(x = "Cluster", y = "Proportion", fill = "Diabetes Status") +

ggtitle("Distribution of Diabetes Status by Cluster (Proportion by Mean Outcome)")</pre>
```

Distribution of Diabetes Status by Cluster (Proportion by Mean Outcome)



```
# Plot the relationship between glucose and insulin, colored by cluster
ggplot(df, aes(x = glucose, y = insulin, color = factor(cluster))) +
geom_point() +
labs(x = "Glucose", y = "Insulin", color = "Cluster") +
ggtitle("Relationship between Glucose and Insulin by Cluster")
```

Relationship between Glucose and Insulin by Cluster

