# Event Data Analysis

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### **Loading Packages**

```
library(dplyr)
library(tidyverse)
library(stringr)
library(lubridate)
library(tidyr)
library(forcats)
library(ggplot2)
library(viridis)
library(viridisLite)
library(purrr)
library(gganimate)
library(ggthemes)
library(leaflet)
library(patchwork)
library(readr)
library(janitor)
library(plotly)
```

#### Data Wrangling

```
# Define the directory containing the files
data_dir <- "data/"</pre>
# Define the file codes
file_codes <- c("F15", "W16", "S16", "F16", "W17", "S17", "F17", "W18", "S18", "F18",
                 "W19", "S19", "F19", "W20", "S20", "F20", "W21", "S21", "F21", "W22",
                 "S22", "F22", "W23", "S23", "F23", "W24", "S24")
# Create a function to read a file given its code
read_and_clean_event_file <- function(code) {</pre>
  file_path <- pasteO(data_dir, "2015-2024 Events Data - ", code, " - Event Data.csv")</pre>
  data <- read csv(file path)</pre>
  clean_names(data) %>%
    mutate(across(everything(), as.character)) %>%
    mutate(term = code)
}
term_to_year <- function(term) {</pre>
  year <- as.numeric(str_sub(term, 2, 3))</pre>
  season <- str_sub(term, 1, 1)</pre>
```

```
start_year <- if_else(season == "F", 2000 + year, 2000 + year - 1)</pre>
  end_year <- start_year + 1</pre>
  return(paste0(start_year, "-", end_year))
# Use purrr to read all files and store them in a named list
event_data_list <- set_names(map(file_codes, read_and_clean_event_file), file_codes)</pre>
combined_data <- reduce(event_data_list, full_join)</pre>
combined_data_filtered <- combined_data %>%
  filter(!is.na(what), what != "") %>%
  filter(what != "Choir & Jazz Rehearsal") %>%
  filter(what != "Jazz Rehearsal") %>%
  mutate(date = as.Date(ymd(date))) %>%
  mutate(support_level = if_else(support_level == "N" | support_level == "Y", "L", support_level)) %%
  mutate(
   department = str_replace_all(department, "WCC", "ODOA"),
    department = str_replace_all(department, "MSUC", "MUSC"),
    department = str_replace_all(department, "French Dept|French", "FREN"),
   department = str_replace_all(department, "English", "ENGL"),
   department = str_replace_all(department, "Pres. Office", "PRES"),
    department = str_replace_all(department, "History", "HIST"),
    department = str_replace_all(department, "THD", "THDA"),
    department = str_replace_all(department, "Inclusion & Equity", "IEC"),
   venue = str_replace_all(venue, "Skinner Chapel", "Chapel"),
   department = str_replace_all(department, "/", " & "),
    department = str_replace_all(department, ",", " &"),
   venue = str_replace_all(venue, ",", " &"),
    department = str_replace_all(department, "\\s+", " ") # Remove extra spaces
  ) %>%
  select(-wk) %>%
  mutate(department_type = case_when(
      department == "MUSC" ~ "MUSC",
      department == "ODOA" ~ "ODOA",
     department == "CSA" ~ "CSA",
      str_detect(department, "&") ~ "Collab",
     TRUE ~ "Others"
   )) %>%
  mutate(what = str_replace_all(what, "Jazz Ensemble Concert|Jazz Area Concert", "Jazz Concert"),
         what = str_replace_all(what, "Symphony Band Concert", "Symphony Concert"),
         what = str_replace_all(what, "Composition Recital", "Composition Showcase Recital"),
         what = str_replace_all(what, "Harpichord", "Harpsichord"),
         what = str_replace_all(what, "Emsemble", "Ensemble"),
         what = str_replace_all(what, "Juest Cellin'", "Just Cellin'"),
         what = str_replace_all(what, "Facutly|FACULTY|Mazariello", "Faculty")) %>%
  mutate(event_type = case_when(
   str_detect(what, "GUEST|ODOA|Concert Series|SPCO") ~ "Guest",
    str_detect(what, "Faculty") ~ "Faculty Recital",
   str_detect(what, "Student|Senior|Junior|Piano Recital: |Johnson|Verma Jameson") ~ "Student Recital"
   str_detect(what, "Studio Recital|Organ & Harpsichord|Composition Showcase Recital|Chamber Recital|C
   str_detect(what, "Orchestra Concert|Jazz Concert|Symphony Concert|Symphony Band|Choir Concert|Orche
    str_detect(what, "CSA|Just Cellin|Lunar New Year|ACA|A Cappella|Accidentals|Exit 69|Date Knight|Kni
```

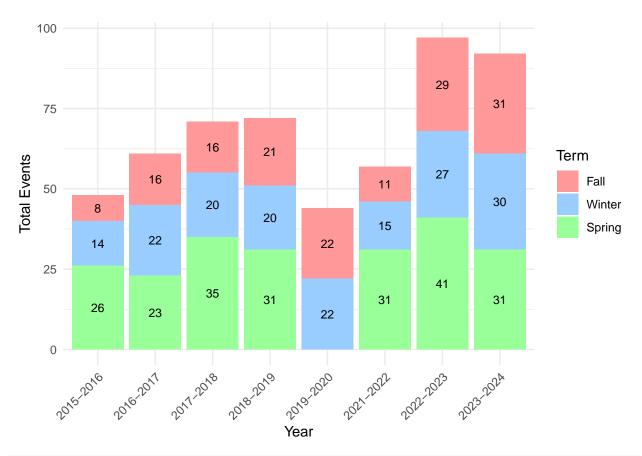
```
str_detect(what, "Masterclass|Lecture|Symposium") ~ "Masterclass",
   str_detect(what, "Trustees|Trustee's|Presidents|Conference|President's|Presentation") ~ "Presentati
   str_detect(what, "Clinic|Music Fest|Music Department Showcase|Melinda Russell|Launch|Event|Opening"
    TRUE ~ "Guest"
 )) %>%
 mutate(year = term_to_year(term)) %>%
 mutate(term = factor(term, levels = c("F15", "W16", "S16", "F16", "W17", "S17",
                              "F17", "W18", "S18", "F18", "W19", "S19",
                               "F19", "W20", "S20", "F20", "W21", "S21",
                               "F21", "W22", "S22", "F22", "W23", "S23",
                              "F23", "W24", "S24"), ordered = TRUE)) %>%
 arrange(year, term) %>%
 mutate(
  term_category = case_when(
    str_detect(term, "^F") ~ "Fall",
    str_detect(term, "^W") ~ "Winter",
    str_detect(term, "^S") ~ "Spring"
  )
 ) %>%
 mutate(term_category = factor(term_category, levels = c("Spring", "Winter", "Fall"), ordered = TRUE))
glimpse(combined_data_filtered)
Rows: 542
Columns: 24
$ venue
             <chr> "Concert Hall", "Concert Hall", "Concert Hall", "Concer
$ date
             <date> 2015-10-25, 2015-10-30, 2015-11-07, 2015-11-08, 2015-~
$ what
             <chr> "Jazz Concert", "Symphony Concert", "Choir Concert", "~
             <chr> "MUSC", "MUSC", "MUSC", "MUSC", "CSA", "MUSC",~
$ department
$ av_staff
             $ pac_staff
             <chr> "H", "M", "M", "H", "H", "L", "M", "L", "M", "L", "L", "
$ support_level
             $ audio_needs
$ stage_needs
             <chr> "H", "H", "H", "H", "H", "M", "M", NA, "M", "L", "L", ~
<chr> "N", "Y", "N", "Y", "N", "N", NA, "N", "N", "N", ~
$ projection
$ video_recording <chr> "N", "N", "Y", "Y", "Y", "Y", "N", NA, "Y", "N", "Y", ~
             <chr> "N", "N", "N", "N", "N", "N", "N", NA, NA, NA, NA, NA
$ live_stream
            <chr> "64", "122", "86", "77", "220", "192", "44", NA, "137"~
$ audience_count
$ poster
             $ program
             $ reception
             $ term
             $ livestream
$ department_type <chr> "MUSC", "MUSC", "MUSC", "MUSC", "MUSC", "CSA", "MUSC",~
             <chr> "Ensemble Concert", "Ensemble Concert", "Ensemble Conc~
$ event_type
             <chr> "2015-2016", "2015-2016", "2015-2016", "2015-2016", "2~
$ year
$ term_category
             <ord> Fall, Fall, Fall, Fall, Fall, Fall, Fall, Fall, Winter~
event_summary <- combined_data_filtered %>%
 group_by(year, term) %>%
 summarize(term_total = n(), .groups = 'drop') %>%
```

```
group_by(year) %>%
mutate(year_total = sum(term_total)) %>%
ungroup()
```

#### **EDA Plots**

### **Overall Event Summary**

```
# Sort the data frame by year and term in the order of F, W, S
event summary <- event summary %>%
 mutate(term = factor(term, levels = c("F15", "W16", "S16", "F16", "W17", "S17",
                                        "F17", "W18", "S18", "F18", "W19", "S19",
                                        "F19", "W20", "S20", "F20", "W21", "S21",
                                        "F21", "W22", "S22", "F22", "W23", "S23",
                                        "F23", "W24", "S24"), ordered = TRUE)) %>%
  arrange(year, term) %>%
  mutate(
   term_category = case_when(
     str_detect(term, "^F") ~ "Fall",
     str_detect(term, "^W") ~ "Winter",
     str_detect(term, "^S") ~ "Spring"
  ) %>%
  mutate(term_category = factor(term_category, levels = c("Fall", "Winter", "Spring"), ordered = TRUE))
# Create the stacked bar chart
ggplot(event_summary, aes(x = year, y = term_total, fill = term_category)) +
  geom_bar(stat = "identity") +
  geom_text(aes(label = term_total),
           position = position_stack(vjust = 0.5),
            size = 3,
            color = "black") +
  scale_fill_manual(values = c("Fall" = "#FF9999", "Winter" = "#99CCFF", "Spring" = "#99FF99")) +
  labs(x = "Year", y = "Total Events", fill = "Term") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

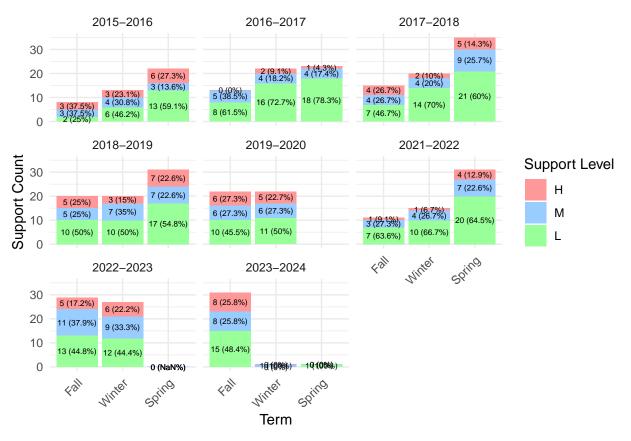


```
event summary
# A tibble: 23 x 5
   year
             term term_total year_total term_category
   <chr>
             <ord>
                         <int>
                                    <int> <ord>
 1 2015-2016 F15
                            8
                                       48 Fall
 2 2015-2016 W16
                                       48 Winter
                            14
3 2015-2016 S16
                            26
                                       48 Spring
 4 2016-2017 F16
                            16
                                       61 Fall
5 2016-2017 W17
                            22
                                       61 Winter
6 2016-2017 S17
                            23
                                       61 Spring
7 2017-2018 F17
                                       71 Fall
                            16
8 2017-2018 W18
                            20
                                       71 Winter
9 2017-2018 S18
                            35
                                       71 Spring
10 2018-2019 F18
                            21
                                       72 Fall
# i 13 more rows
```

### Breakdown of Events by Support Level

```
library(dplyr)
library(tidyr)
library(purrr)
library(ggplot2)
# Function to summarize and pivot data for each year
```

```
summarize_and_pivot <- function(current_year) {</pre>
  combined_data_filtered %>%
   filter(year == current_year) %>%
    group_by(term_category, support_level) %>%
    summarize(Support = n(), .groups = 'drop') %>%
   pivot_wider(names_from = support_level, values_from = Support, values_fill = list(Support = 0)) %>%
   mutate(year = current_year) # Add a year column to identify the table
}
# List of years to iterate over
years <- unique(combined_data_filtered$year)</pre>
# Use purrr::map to apply summarize_and_pivot function for each year
support_tables <- map_dfr(years, summarize_and_pivot) # Combine into a single data frame
# Melt the data for ggplot
support_tables_melted <- support_tables %>%
  pivot_longer(cols = -c(term_category, year), names_to = "support_level", values_to = "Support")
# Ensure the term_category and support_level are factors with the correct order
support_tables_melted <- support_tables_melted %>%
  filter(support_level != "NA") %>%
  mutate(term_category = factor(term_category, levels = c("Fall", "Winter", "Spring"), ordered = TRUE),
         support_level = factor(support_level, levels = c("NA", "H", "M", "L"), ordered = TRUE))
# Calculate percentages for each segment
support_tables_melted <- support_tables_melted %>%
  group_by(year, term_category) %>%
  mutate(total_support = sum(Support),
         percentage = (Support / total_support) * 100)
# Create the facet plot
ggplot(support_tables_melted, aes(x = term_category, y = Support, fill = support_level)) +
  geom_bar(stat = "identity", position = "stack") +
  geom_text(aes(label = paste0(Support, " (", round(percentage, 1), "%)")),
            position = position_stack(vjust = 0.5), size = 2, color = "black") +
 facet_wrap(~ year) +
  labs(x = "Term", y = "Support Count", fill = "Support Level") +
  scale_fill_manual(values = c("H" = "#FF9999", "M" = "#99CCFF", "L" = "#99FF99", "NA" = "black")) +
  theme minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

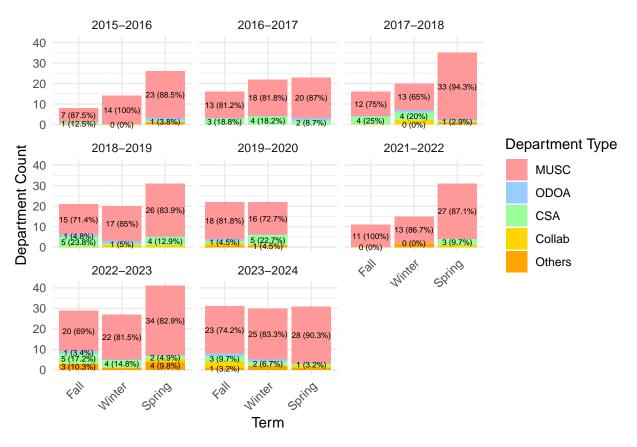


```
support tables melted
# A tibble: 69 x 6
# Groups:
            year, term_category [23]
   term_category year
                            support_level Support total_support percentage
   <ord>
                            <ord>
                                             <int>
                                                            <int>
                                                                        <dbl>
                  <chr>
                                                                        27.3
 1 Spring
                  2015-2016 H
                                                  6
                                                                22
                                                                        59.1
                  2015-2016 L
                                                 13
                                                                22
 2 Spring
                  2015-2016 M
                                                  3
                                                                22
                                                                        13.6
 3 Spring
 4 Winter
                  2015-2016 H
                                                  3
                                                                13
                                                                        23.1
 5 Winter
                  2015-2016 L
                                                  6
                                                               13
                                                                        46.2
                  2015-2016 M
 6 Winter
                                                  4
                                                                13
                                                                        30.8
 7 Fall
                  2015-2016 H
                                                  3
                                                                8
                                                                        37.5
                                                  2
8 Fall
                  2015-2016 L
                                                                8
                                                                        25
                                                  3
                                                                8
                                                                        37.5
9 Fall
                  2015-2016 M
10 Spring
                  2016-2017 H
                                                                23
                                                                         4.35
# i 59 more rows
```

## Breakdown of Events by Department/Source

```
# Function to summarize and pivot data for each year
summarize_and_pivot_department <- function(current_year) {
  combined_data_filtered %>%
    filter(year == current_year) %>%
    group_by(term_category, department_type) %>%
    summarize(DepartmentCount = n(), .groups = 'drop') %>%
```

```
pivot_wider(names_from = department_type, values_from = DepartmentCount, values_fill = list(Department_type)
   mutate(year = current_year) # Add a year column to identify the table
}
# List of years to iterate over
years <- unique(combined_data_filtered$year)</pre>
# Use purrr::map to apply summarize_and_pivot function for each year
department_tables <- map_dfr(years, summarize_and_pivot_department) # Combine into a single data frame
# Melt the data for ggplot
department_tables_melted <- department_tables %>%
  pivot longer(cols = -c(term category, year), names to = "department type", values to = "DepartmentCou
# Ensure the term_category and department_type are factors with the correct order
department_tables_melted <- department_tables_melted %>%
  mutate(term_category = factor(term_category, levels = c("Fall", "Winter", "Spring"), ordered = TRUE),
         department_type = factor(department_type, levels = c("MUSC", "ODOA", "CSA", "Collab", "Others"
# Filter out NA department types (if any)
department_tables_melted <- department_tables_melted %>%
  filter(department_type != "NA")
# Calculate percentages for each segment
department_tables_melted <- department_tables_melted %>%
  group_by(year, term_category) %>%
  mutate(total_count = sum(DepartmentCount, na.rm = T),
         percentage = (DepartmentCount / total_count) * 100)
# Create the facet plot
ggplot(department_tables_melted, aes(x = term_category, y = DepartmentCount, fill = department_type)) +
  geom_bar(stat = "identity", position = "stack") +
  geom_text(aes(label = paste0(DepartmentCount, " (", round(percentage, 1), "%)")),
            position = position_stack(vjust = 0.5), size = 2, color = "black",
            check_overlap = TRUE) +
  facet_wrap(~ year) +
  labs(x = "Term", y = "Department Count", fill = "Department Type") +
  scale_fill_manual(values = c("MUSC" = "#FF9999", "ODOA" = "#99CCFF", "CSA" = "#99FF99", "Collab" = "#
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

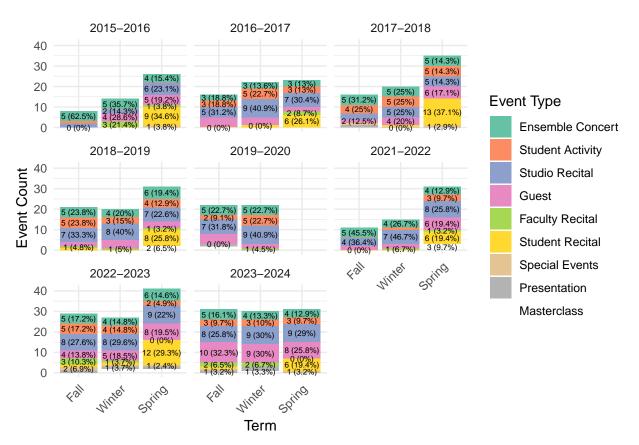


```
department tables melted
# A tibble: 115 x 6
# Groups:
            year, term_category [23]
   term_category year
                           department_type DepartmentCount total_count percentage
   <ord>
                           <ord>
                                                                    <int>
                                                                                <dbl>
                  <chr>
                                                       <int>
                                                                                 3.85
 1 Spring
                  2015-20~ CSA
                                                                       26
                  2015-20~ MUSC
                                                          23
                                                                                88.5
 2 Spring
                                                                       26
                  2015-20~ ODOA
                                                                       26
                                                                                 3.85
 3 Spring
                                                           1
 4 Spring
                  2015-20~ Others
                                                                       26
                                                                                 3.85
 5 Spring
                  2015-20~ Collab
                                                          NA
                                                                       26
                                                                                NA
 6 Winter
                  2015-20~ CSA
                                                           0
                                                                       14
                                                                                 0
                  2015-20~ MUSC
                                                          14
                                                                       14
                                                                               100
 7 Winter
 8 Winter
                  2015-20~ ODOA
                                                           0
                                                                       14
                                                                                 0
                  2015-20~ Others
                                                                                 0
9 Winter
                                                           0
                                                                       14
10 Winter
                  2015-20~ Collab
                                                                       14
                                                                                NA
                                                          NA
# i 105 more rows
```

## Breakdown of Music, Collab & ODOA Events by Type

```
# Function to summarize and pivot data for each year
summarize_and_pivot_event <- function(current_year) {
  combined_data_filtered %>%
    filter(year == current_year) %>%
    group_by(term_category, event_type) %>%
    summarize(EventCount = n(), .groups = 'drop') %>%
```

```
pivot_wider(names_from = event_type, values_from = EventCount, values_fill = list(EventCount = 0))
   mutate(year = current_year) # Add a year column to identify the table
}
# List of years to iterate over
years <- unique(combined_data_filtered$year)</pre>
# Use purrr::map to apply summarize and pivot function for each year
event_tables <- map_dfr(years, summarize_and_pivot_event) # Combine into a single data frame
# Melt the data for ggplot
event_tables_melted <- event_tables %>%
  pivot_longer(cols = -c(term_category, year), names_to = "event_type", values_to = "EventCount")
# Ensure the term_category and event_type are factors with the correct order
event_tables_melted <- event_tables_melted %>%
  mutate(term_category = factor(term_category, levels = c("Fall", "Winter", "Spring"), ordered = TRUE),
         event_type = factor(event_type, levels = c("Ensemble Concert", "Student Activity", "Studio Rec
                                                    "Guest", "Faculty Recital", "Student Recital",
                                                    "Special Events", "Presentation", "Masterclass"),
                            ordered = TRUE))
# Calculate percentages for each segment
event_tables_melted <- event_tables_melted %>%
  group by(year, term category) %>%
  mutate(total_count = sum(EventCount, na.rm = T),
         percentage = (EventCount / total_count) * 100)
# Create the facet plot
ggplot(event_tables_melted, aes(x = term_category, y = EventCount, fill = event_type)) +
  geom_bar(stat = "identity", position = "stack") +
  geom_text(aes(label = paste0(EventCount, " (", round(percentage, 1), "%)")),
            position = position_stack(vjust = 0.5), size = 2, color = "black",
            check_overlap = TRUE) +
  facet_wrap(~ year) +
  labs(x = "Term", y = "Event Count", fill = "Event Type") +
  scale_fill_brewer(palette = "Set2") + # Using a Brewer palette for default colors
  theme minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



A tibble: 207	ted x 6				
Groups: year	, term_cate	egory [23]			
term_category	year	event_type	${\tt EventCount}$	total_count	percentage
<ord></ord>	<chr></chr>	<ord></ord>	<int></int>	<int></int>	<dbl></dbl>
1 Spring	2015-2016	Ensemble Concert	4	26	15.4
2 Spring	2015-2016	Faculty Recital	1	26	3.85
3 Spring	2015-2016	Guest	5	26	19.2
Spring	2015-2016	Special Events	1	26	3.85
Spring	2015-2016	Student Recital	9	26	34.6
Spring	2015-2016	Studio Recital	6	26	23.1
Spring	2015-2016	Student Activity	0	26	0
Spring	2015-2016	Presentation	NA	26	NA
Spring	2015-2016	Masterclass	NA	26	NA
Winter	2015-2016	Ensemble Concert	5	14	35.7