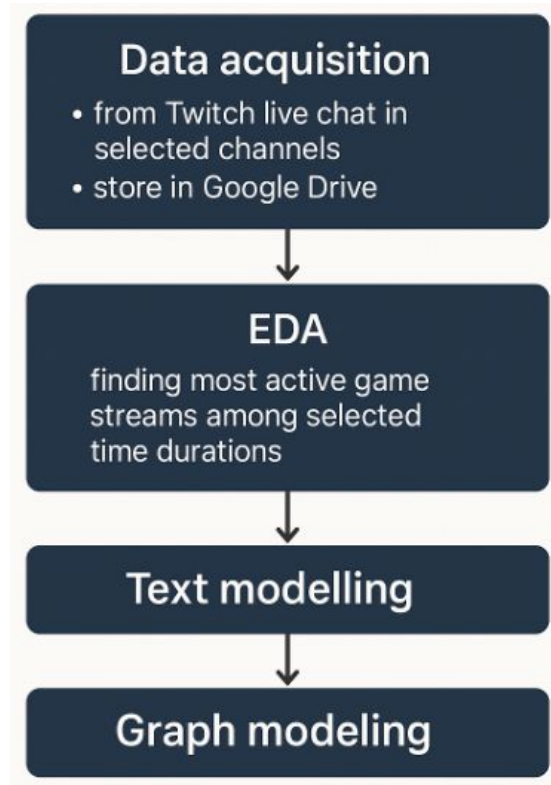


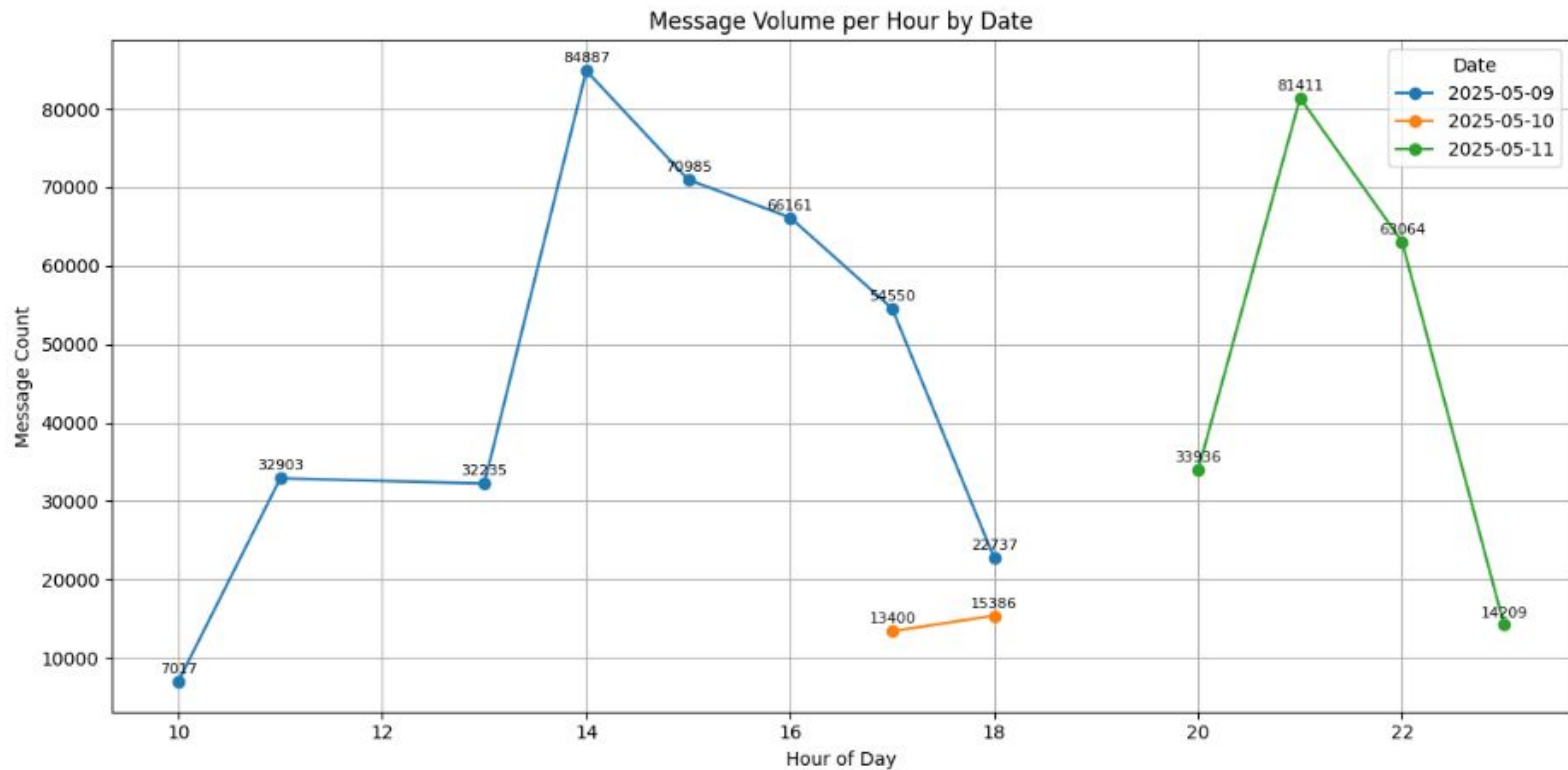
# twitch stream chat modeling

Jia Jia  
Shameek

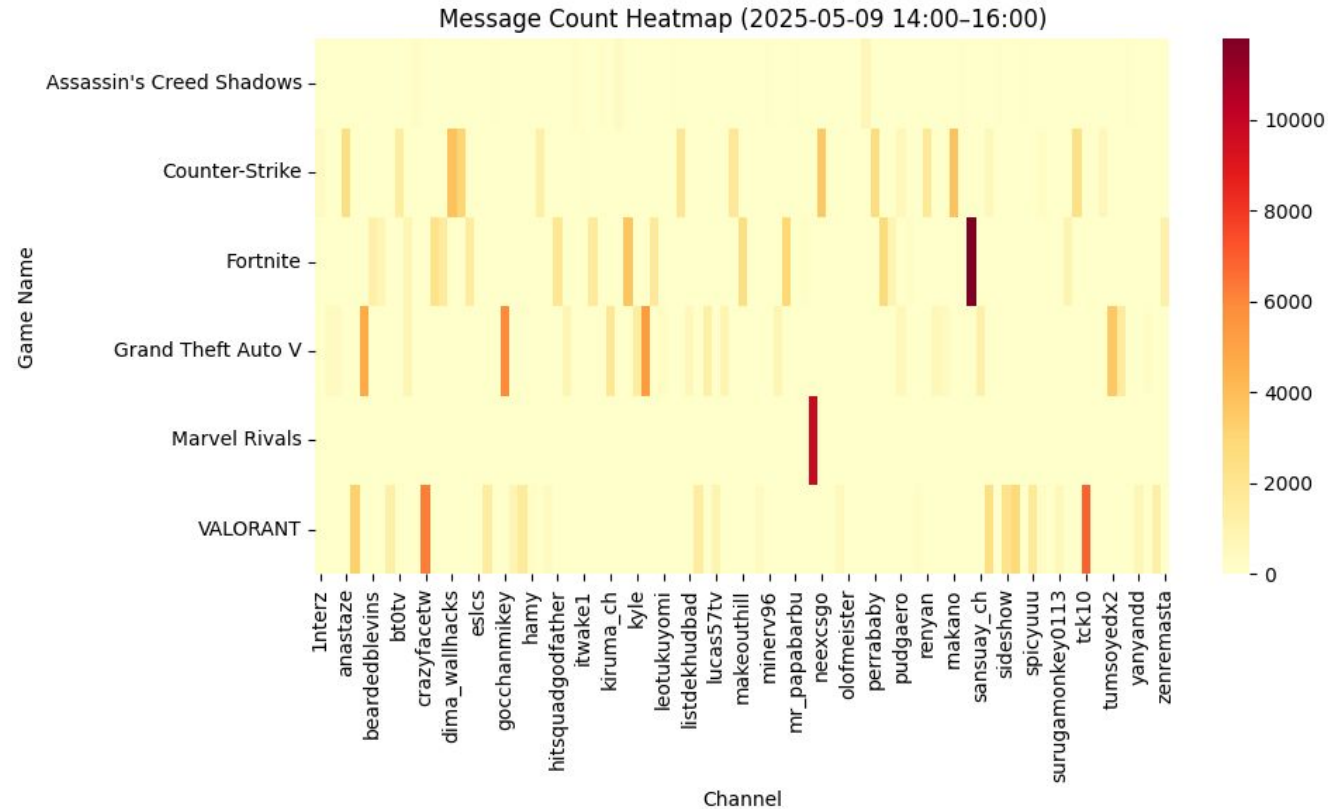
# Data Project Process



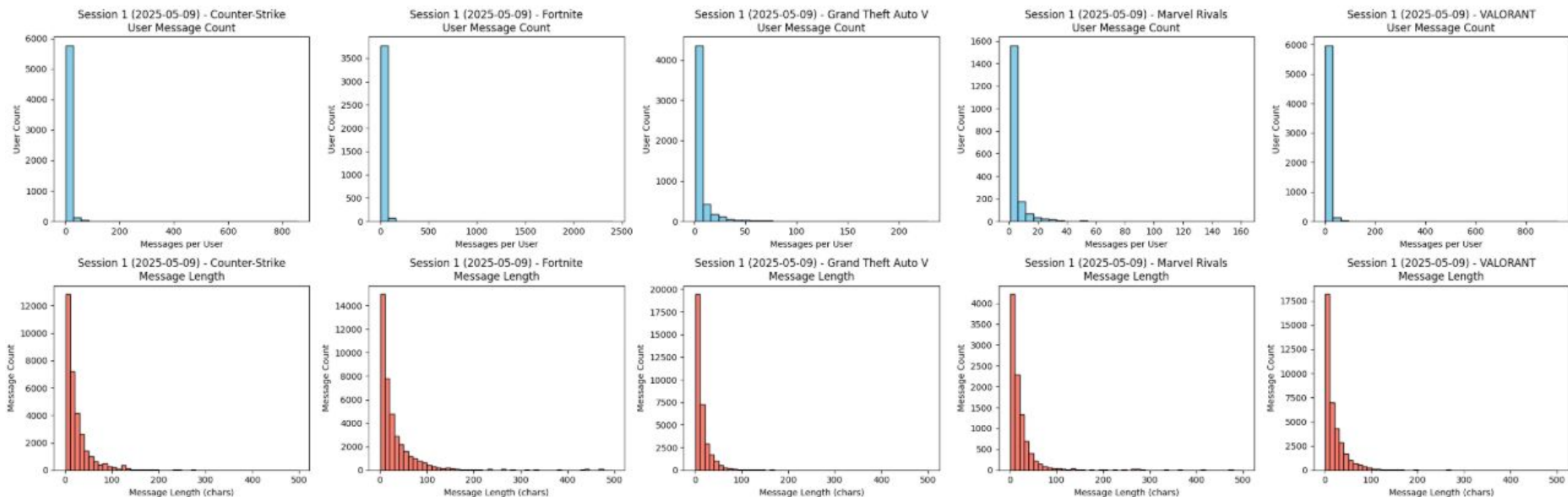
# EDA



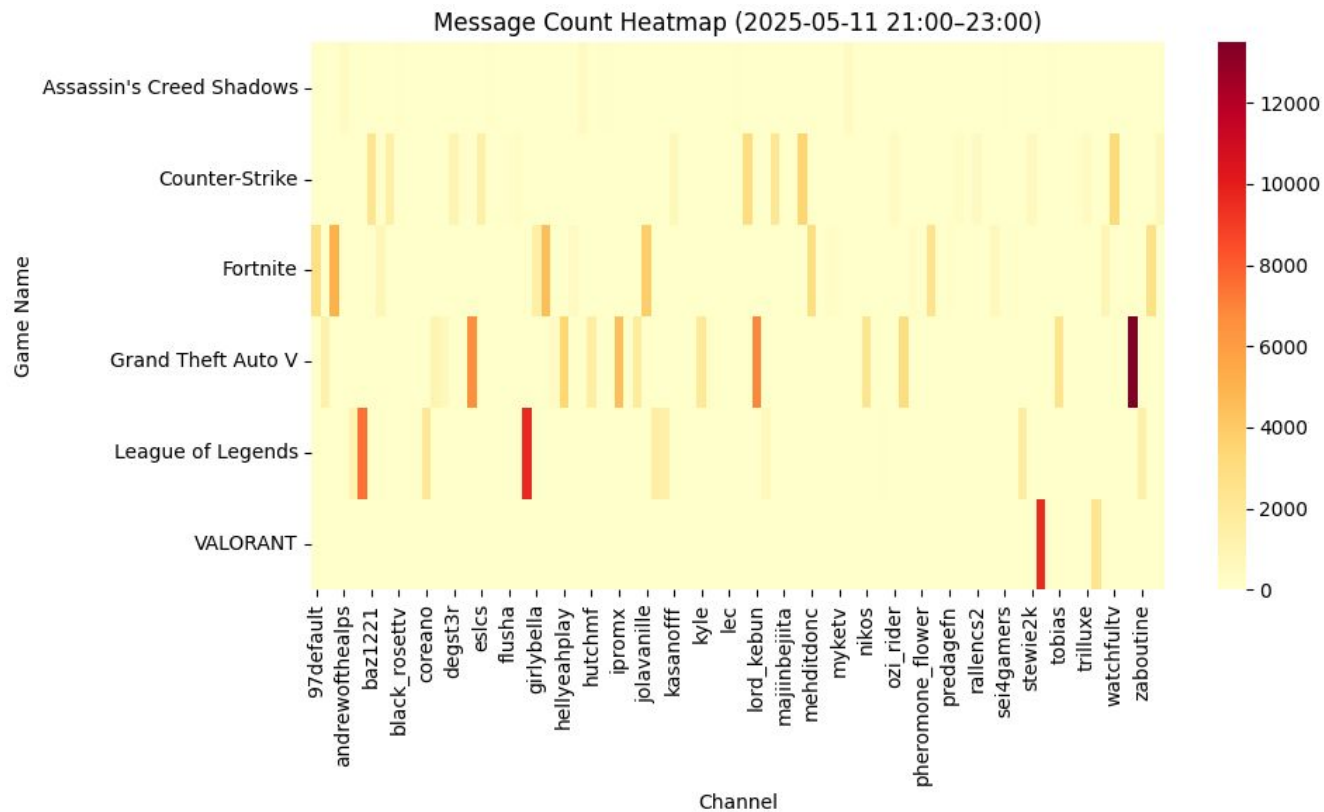
# EDA: session1



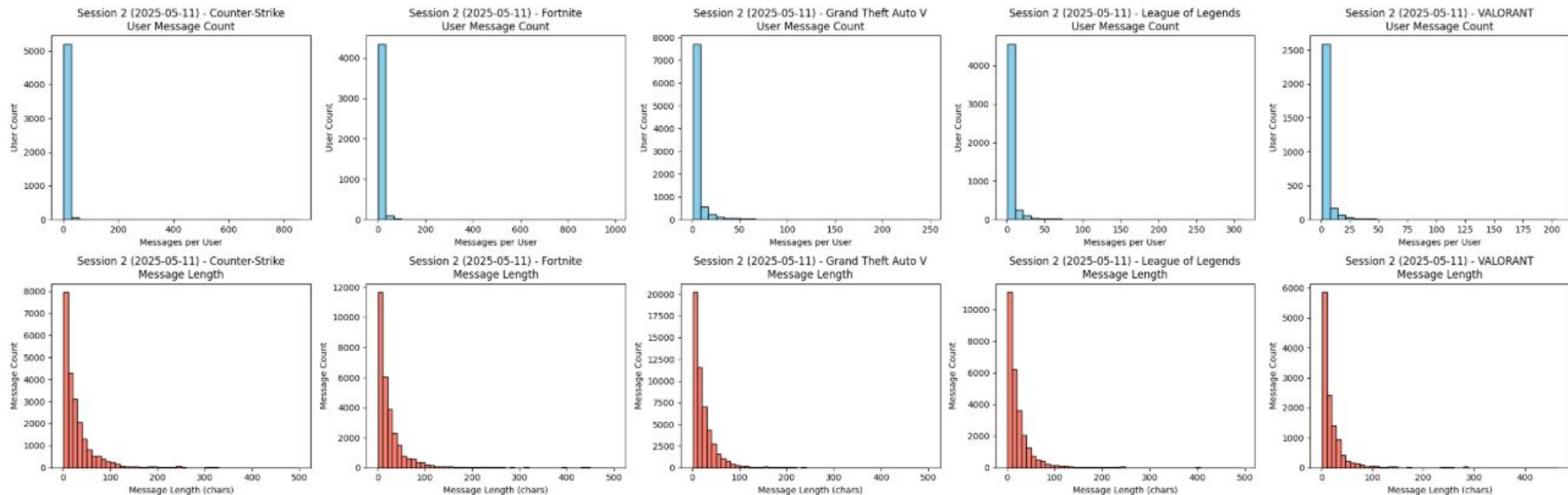
# EDA: session1



# EDA: session2

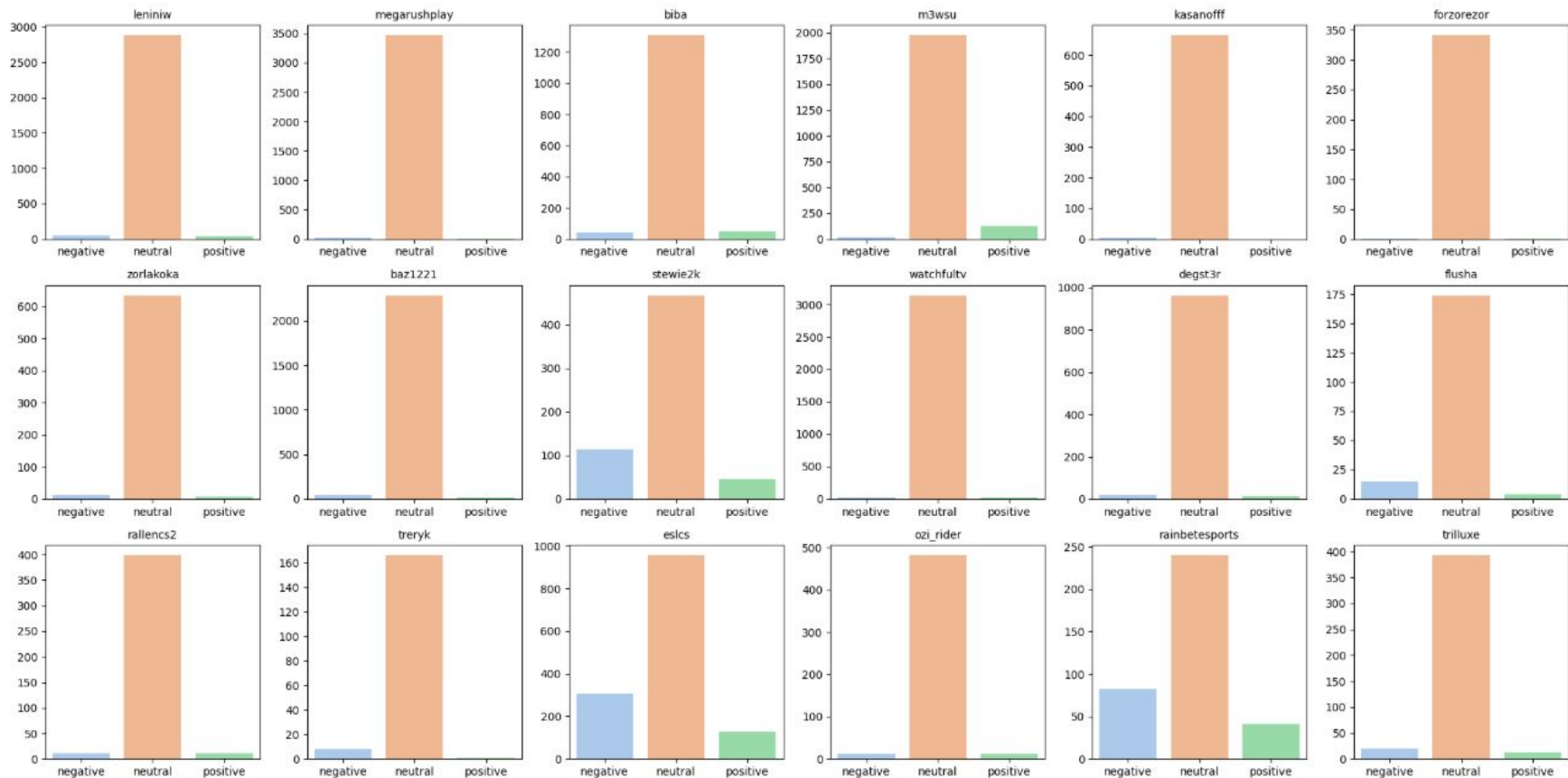


# EDA: session2



# Text Modeling: Counter-Strike in Session2

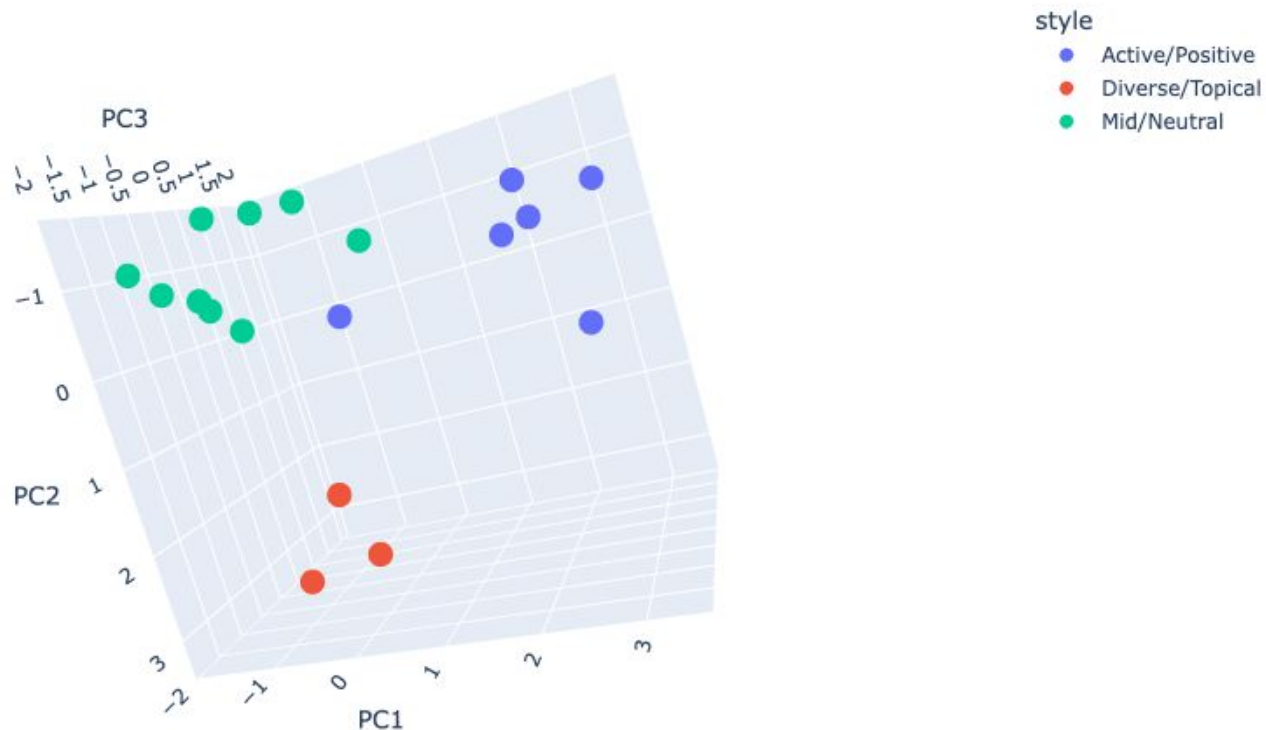
Sentiment Distribution per Channel



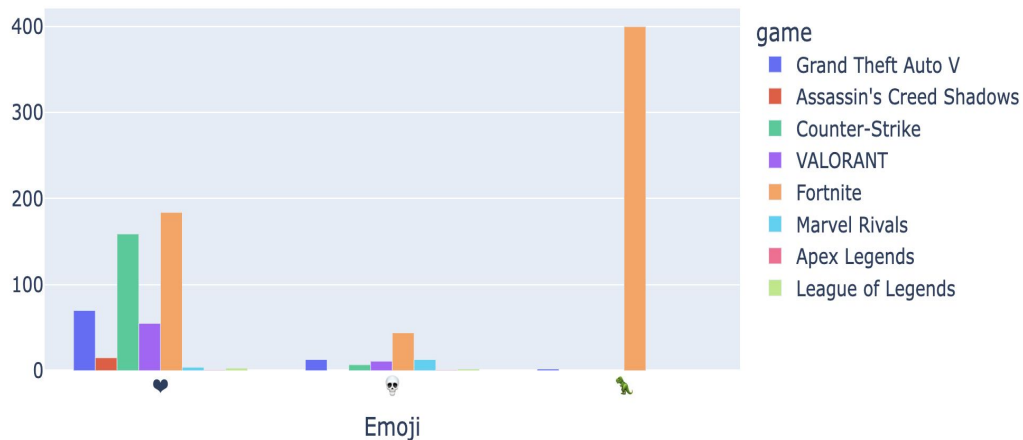


# Text Clustering: Counter-Strike in Session2

🎮 Channel Style Clustering (3D PCA)

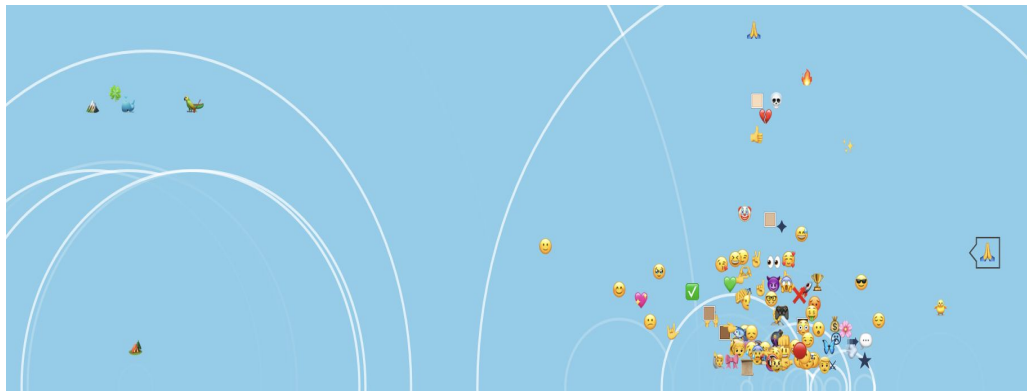


# Emoji Preferences Across Games



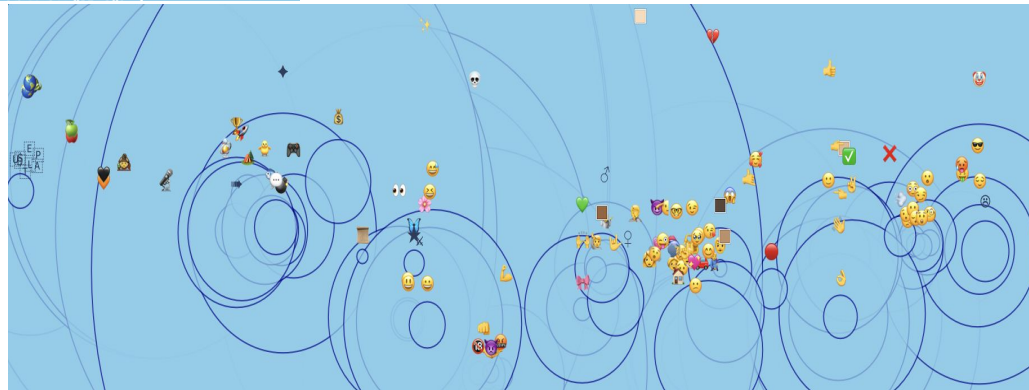
Before diving into semantic embeddings, let's first look at how emoji usage varies by game

# t-SNE of Emoji Embeddings



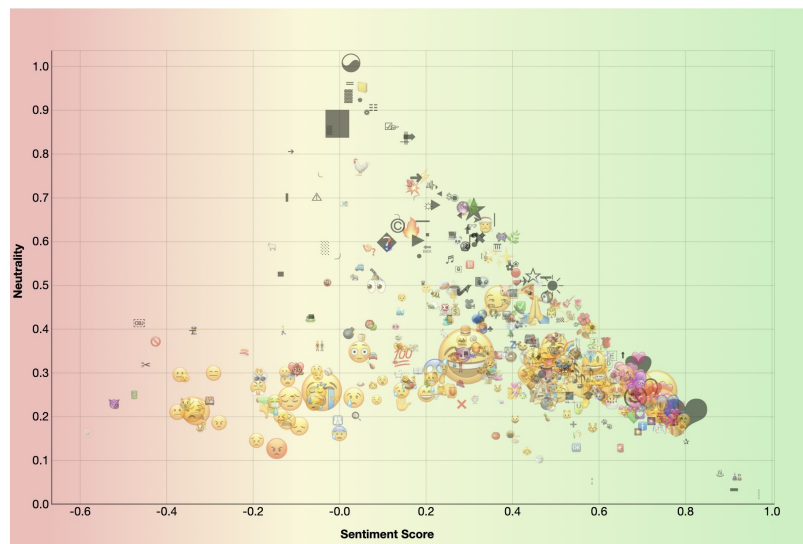
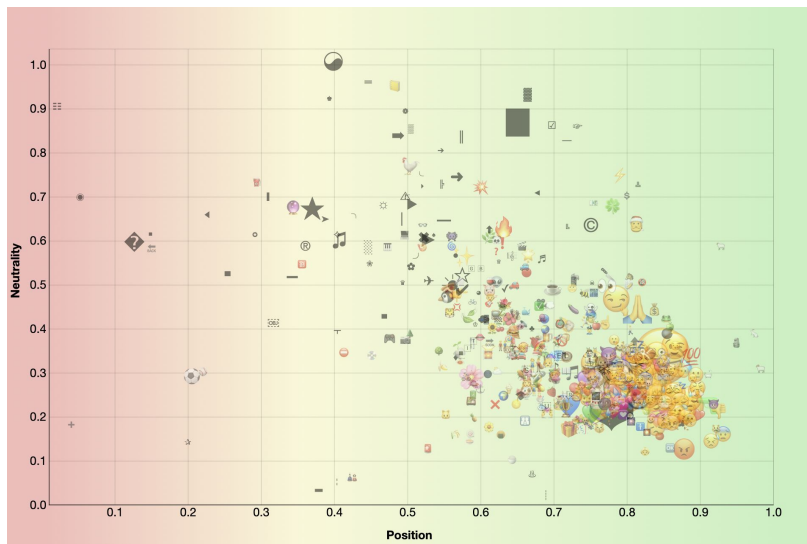
Some dense regions correspond to tactical game callouts, while others are driven by reactions, memes or inside jokes

Even without labels, structure emerges, showing how shared language, repetition, and cultural references organise chat semantically

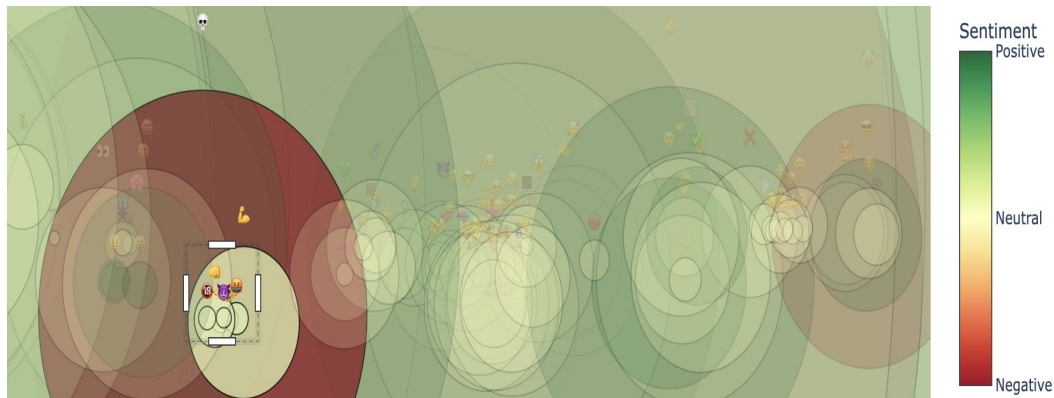
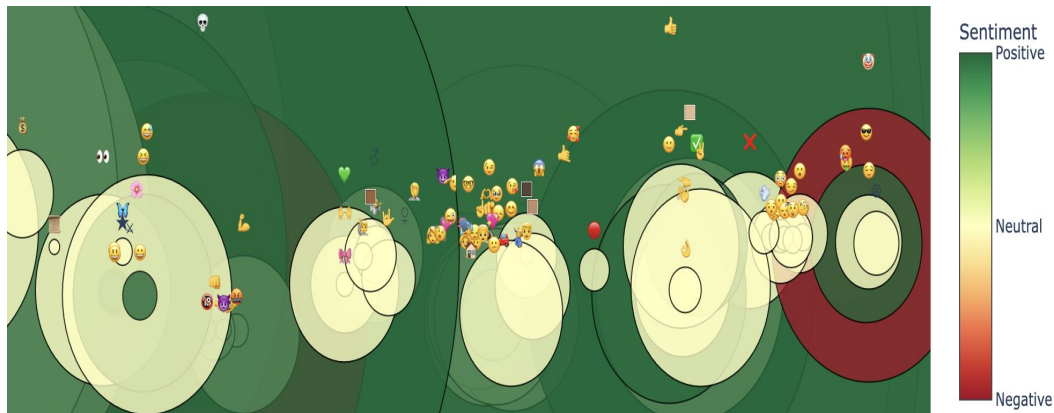


# We move from words to emojis: Emoji Sentiment Score

Using a combination of external sentiment rankings (like Emoji Sentiment Ranking from Škrlj et al.) and custom mappings



# Emoji Co-occurrence Embeddings



Emojis that appear together often are embedded closer using Node2Vec.

We then color them using a sentiment gradient red for negative, green for positive, gray for neutral.

What emerges is a kind of emotional map of chat