### **Bivariate Kernel Density Estimation Report: Titanic Dataset**

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## 1. Data Description and Source

The dataset used in this analysis is the **Titanic training dataset**, obtained from the Kaggle Titanic: Machine Learning from Disaster competition. Link to the data set: (https://www.kaggle.com/competitions/titanic/data)

It contains detailed information on passengers aboard the Titanic, including these variables:

- survival(survival)
- pclass (ticket class)
- sex(sex)
- age(age in years)
- sibsp(of siblings / spouses aboard the Titanic)
- parch(of parents / children aboard the Titanic)
- ticket( ticket number), fare(passenger fare)
- cabin(cabin number)
- embarked (port of embarkation).

For the purpose of this kernel density analysis, two continuous variables were selected:

**Age**: Passenger's age in years

Fare: Ticket price in British Pounds

Rows with missing values in either Age or Fare were excluded to ensure a complete-case analysis, resulting in a clean dataset suitable for bivariate density estimation.

## 2. Plots of the Bivariate Density Distribution

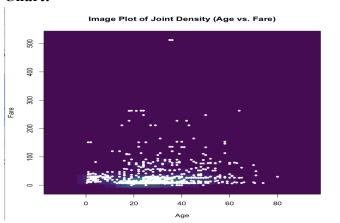
To understand the joint distribution of Age and Fare, we used the ks package in R to estimate the **bivariate kernel density** then the results are visualized

# **◆** Image (Heat) Plot

```
# b. Visualize using an Image (Heat) Plot with a Viridis Color Palette
image(x = kde_result$eval.points[[1]],
        y = kde_result$eval.points[[2]],
        z = kde_result$estimate,
        col = viridis(20),
        xlab = "Age",
        ylab = "Fare",
        main = "Image Plot of Joint Density (Age vs. Fare)")

# Overlay the original data points to add_context
points(kde_result$x, pch = 20, col = "white")
```

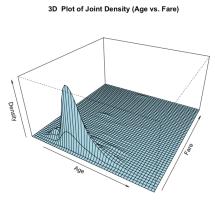
#### **Chart:**



- The overall distribution is **right-skewed** on the Fare axis.
- There's **no strong linear relationship** between Age and Fare.
- Most passengers were **young to middle-aged** and traveled on **budget fares**, with a **concentrated cluster** visible in the lower-left region of the plot.

# **◆ 3D Perspective Plot**

A three-dimensional surface plot was created to illustrate the topology of the density landscape.



- The majority of passengers were younger and paid cheaper fares.
- There's **no strong dependency** between Age and Fare—high-density areas are localized rather than trending diagonally.
- This 3D view reinforces the findings from the heat plot: the data is **skewed toward low-fare**, **younger individuals**, with limited density elsewhere.

#### Slice and Contour Plots

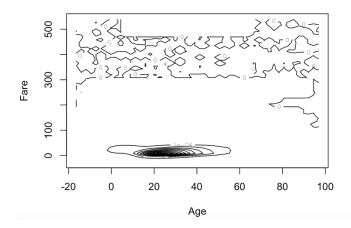
```
#Slice plot
plot(kde_result, display="filled.contour2")
```

000 -20 0 20 40 60 80 100 x[, 1]

- The contour plot reaffirms that most passengers were **young to middle-aged adults** who paid **low to moderate fares**.
- The distribution is **unimodal** and fairly compact, with **no strong nonlinear patterns**.
- Age and Fare are **not strongly correlated**, but they do cluster in one dominant region.

#### Contour Plot

### Contour Plot of Joint Density (Age vs. Fare)



- The contour plot confirms that the **majority of Titanic passengers were younger** and paid lower fares.
- The distribution is **unimodal** with a compact high-density core.
- Age and Fare are not strongly correlated; the relationship appears weak, with density driven more by passenger clustering in certain demographics.

## 3. Conclusions: Shape, Direction, and Strength

The joint distribution is heavily **right-skewed** on the Fare axis on all plot. Most passengers paid lower fares and fell within a broad age range of 20–40.

**Direction**: No strong **linear relationship** exists between Age and Fare. The plot shows clusters of younger passengers at low fares and some older passengers paying both low and high fares (likely due to first-class status).

**Strength**: The relationship between Age and Fare is **weak**. The density contours are spread out with no clear trend, suggesting little to no correlation between these two variables.

#### 4. Bandwidth Matrix HH

The bandwidth matrix HH determines the level of smoothing in kernel density estimation. In this analysis, we computed HH using **smoothed cross-validation** via the Hscv() function from the **ks** package. This method selects an optimal bandwidth that balances bias and variance by minimizing mean integrated squared error (MISE).

The resulting HH matrix is symmetric and positive definite, ensuring a smooth and stable estimate of the joint density function. Choosing a good HH is essential:

**Too small** → noisy, overfitted estimate

**Too large** → oversmoothed, loss of structure

In this case, the automatic selection resulted in visually meaningful plots that reflected key patterns in the data.