

FDA_Project

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```
#install.packages("forecast")
```

```
# Load the dataset
```

```
data <- read.csv("D:/Aathmika Vijay/7th semester/foundations of data analytics/Gender Inequality Index.csv")  
head(data,6)
```

##	ISO3	Country	Continent	Hemisphere
## 1	AFG	Afghanistan	Asia	Northern Hemisphere
## 2	AGO	Angola	Africa	Southern Hemisphere
## 3	ALB	Albania	Europe	Northern Hemisphere
## 4	AND	Andorra	Europe	Northern Hemisphere
## 5	ARE	United Arab Emirates	Asia	Northern Hemisphere
## 6	ARG	Argentina	America	Southern Hemisphere
##	Human.Development.Groups UNDP.Developing.Regions HDI.Rank..2021.			
## 1	Low		SA	180
## 2	Medium		SSA	148
## 3	High		ECA	67
## 4	Very High			40
## 5	Very High		AS	26
## 6	Very High		LAC	47
##	GII.Rank..2021. Gender.Inequality.Index..1990. Gender.Inequality.Index..1991.			
## 1	167		NA	NA
## 2	136		0.725	0.723
## 3	39		NA	NA
## 4	NA		NA	NA
## 5	11		0.659	0.647
## 6	69		0.442	0.439
##	Gender.Inequality.Index..1992. Gender.Inequality.Index..1993.			
## 1		NA		NA
## 2		0.721		0.719
## 3		NA		NA
## 4		NA		NA
## 5		0.640		0.632
## 6		0.436		0.430
##	Gender.Inequality.Index..1994. Gender.Inequality.Index..1995.			
## 1		NA		NA
## 2		0.717		0.714
## 3		NA		NA
## 4		NA		NA
## 5		0.635		0.627
## 6		0.429		0.426
##	Gender.Inequality.Index..1996. Gender.Inequality.Index..1997.			
## 1		NA		NA
## 2		0.712		0.708
## 3		NA		NA
## 4		NA		NA
## 5		0.624		0.616
## 6		0.422		0.419
##	Gender.Inequality.Index..1998. Gender.Inequality.Index..1999.			
## 1		NA		NA
## 2		0.683		0.678
## 3		NA		0.330
## 4		NA		NA
## 5		0.609		0.605
## 6		0.415		0.415
##	Gender.Inequality.Index..2000. Gender.Inequality.Index..2001.			
## 1		NA		NA
## 2		0.671		0.665
## 3		0.319		0.282
## 4		NA		NA
## 5		0.605		0.602
## 6		0.413		0.412
##	Gender.Inequality.Index..2002. Gender.Inequality.Index..2003.			

## 1	NA	NA
## 2	0.657	0.649
## 3	0.297	0.305
## 4	NA	NA
## 5	0.599	0.597
## 6	0.390	0.385
## Gender.Inequality.Index..2004.	Gender.Inequality.Index..2005.	
## 1	NA	0.748
## 2	0.643	0.636
## 3	0.301	0.306
## 4	NA	NA
## 5	0.592	0.584
## 6	0.377	0.373
## Gender.Inequality.Index..2006.	Gender.Inequality.Index..2007.	
## 1	0.749	0.752
## 2	0.628	0.621
## 3	0.285	0.286
## 4	NA	NA
## 5	0.311	0.202
## 6	0.369	0.367
## Gender.Inequality.Index..2008.	Gender.Inequality.Index..2009.	
## 1	0.755	0.755
## 2	0.568	0.560
## 3	0.293	0.242
## 4	NA	NA
## 5	0.193	0.185
## 6	0.363	0.369
## Gender.Inequality.Index..2010.	Gender.Inequality.Index..2011.	
## 1	0.753	0.746
## 2	0.551	0.544
## 3	0.246	0.252
## 4	NA	NA
## 5	0.175	0.181
## 6	0.364	0.362
## Gender.Inequality.Index..2012.	Gender.Inequality.Index..2013.	
## 1	0.738	0.728
## 2	0.545	0.540
## 3	0.235	0.225
## 4	NA	NA
## 5	0.171	0.161
## 6	0.360	0.356
## Gender.Inequality.Index..2014.	Gender.Inequality.Index..2015.	
## 1	0.718	0.706
## 2	0.531	0.530
## 3	0.219	0.204
## 4	NA	NA
## 5	0.151	0.126
## 6	0.351	0.345
## Gender.Inequality.Index..2016.	Gender.Inequality.Index..2017.	
## 1	0.692	0.678
## 2	0.529	0.538
## 3	0.191	0.170
## 4	NA	NA
## 5	0.118	0.112
## 6	0.338	0.328
## Gender.Inequality.Index..2018.	Gender.Inequality.Index..2019.	
## 1	0.671	0.665
## 2	0.537	0.537

## 3	0.164	0.156
## 4	NA	NA
## 5	0.103	0.056
## 6	0.315	0.306
## Gender.Inequality.Index..2020.	Gender.Inequality.Index..2021.	
## 1	0.674	0.678
## 2	0.537	0.537
## 3	0.156	0.144
## 4	NA	NA
## 5	0.050	0.049
## 6	0.293	0.287

```
str(data)
```

```
## 'data.frame': 195 obs. of 40 variables:
## $ ISO3 : chr "AFG" "AGO" "ALB" "AND" ...
## $ Country : chr "Afghanistan" "Angola" "Albania" "Andorra" ...
## $ Continent : chr "Asia" "Africa" "Europe" "Europe" ...
## $ Hemisphere : chr "Northern Hemisphere" "Southern Hemisphere" "Northern Hemisp
here" "Northern Hemisphere" ...
## $ Human.Development.Groups : chr "Low" "Medium" "High" "Very High" ...
## $ UNDP.Developing.Regions : chr "SA" "SSA" "ECA" "" ...
## $ HDI.Rank..2021. : int 180 148 67 40 26 47 85 71 5 25 ...
## $ GII.Rank..2021. : int 167 136 39 NA 11 69 53 NA 19 12 ...
## $ Gender.Inequality.Index..1990.: num NA 0.725 NA NA 0.659 0.442 0.47 NA 0.185 0.202 ...
## $ Gender.Inequality.Index..1991.: num NA 0.723 NA NA 0.647 0.439 0.468 NA 0.184 0.204 ...
## $ Gender.Inequality.Index..1992.: num NA 0.721 NA NA 0.64 0.436 0.487 NA 0.184 0.201 ...
## $ Gender.Inequality.Index..1993.: num NA 0.719 NA NA 0.632 0.43 0.491 NA 0.181 0.195 ...
## $ Gender.Inequality.Index..1994.: num NA 0.717 NA NA 0.635 0.429 0.481 NA 0.179 0.18 ...
## $ Gender.Inequality.Index..1995.: num NA 0.714 NA NA 0.627 0.426 0.469 NA 0.176 0.17 ...
## $ Gender.Inequality.Index..1996.: num NA 0.712 NA NA 0.624 0.422 0.463 NA 0.174 0.163 ...
## $ Gender.Inequality.Index..1997.: num NA 0.708 NA NA 0.616 0.419 0.446 NA 0.172 0.155 ...
## $ Gender.Inequality.Index..1998.: num NA 0.683 NA NA 0.609 0.415 0.438 NA 0.169 0.151 ...
## $ Gender.Inequality.Index..1999.: num NA 0.678 0.33 NA 0.605 0.415 0.462 NA 0.156 0.147 ...
## $ Gender.Inequality.Index..2000.: num NA 0.671 0.319 NA 0.605 0.413 0.453 NA 0.151 0.149 ...
## $ Gender.Inequality.Index..2001.: num NA 0.665 0.282 NA 0.602 0.412 0.444 NA 0.15 0.147 ...
## $ Gender.Inequality.Index..2002.: num NA 0.657 0.297 NA 0.599 0.39 0.439 NA 0.144 0.145 ...
## $ Gender.Inequality.Index..2003.: num NA 0.649 0.305 NA 0.597 0.385 0.408 NA 0.14 0.131 ...
## $ Gender.Inequality.Index..2004.: num NA 0.643 0.301 NA 0.592 0.377 0.394 NA 0.136 0.126 ...
## $ Gender.Inequality.Index..2005.: num 0.748 0.636 0.306 NA 0.584 0.373 0.386 NA 0.135 0.121 ...
## $ Gender.Inequality.Index..2006.: num 0.749 0.628 0.285 NA 0.311 0.369 0.376 NA 0.134 0.119 ...
## $ Gender.Inequality.Index..2007.: num 0.752 0.621 0.286 NA 0.202 0.367 0.331 NA 0.134 0.114 ...
## $ Gender.Inequality.Index..2008.: num 0.755 0.568 0.293 NA 0.193 0.363 0.348 NA 0.138 0.121 ...
## $ Gender.Inequality.Index..2009.: num 0.755 0.56 0.242 NA 0.185 0.369 0.341 NA 0.135 0.113 ...
## $ Gender.Inequality.Index..2010.: num 0.753 0.551 0.246 NA 0.175 0.364 0.346 NA 0.136 0.111 ...
## $ Gender.Inequality.Index..2011.: num 0.746 0.544 0.252 NA 0.181 0.362 0.342 NA 0.131 0.103 ...
## $ Gender.Inequality.Index..2012.: num 0.738 0.545 0.235 NA 0.171 0.36 0.327 NA 0.13 0.098 ...
## $ Gender.Inequality.Index..2013.: num 0.728 0.54 0.225 NA 0.161 0.356 0.305 NA 0.123 0.086 ...
## $ Gender.Inequality.Index..2014.: num 0.718 0.531 0.219 NA 0.151 0.351 0.313 NA 0.117 0.083 ...
## $ Gender.Inequality.Index..2015.: num 0.706 0.53 0.204 NA 0.126 0.345 0.309 NA 0.111 0.084 ...
## $ Gender.Inequality.Index..2016.: num 0.692 0.529 0.191 NA 0.118 0.338 0.306 NA 0.101 0.082 ...
## $ Gender.Inequality.Index..2017.: num 0.678 0.538 0.17 NA 0.112 0.328 0.261 NA 0.096 0.072 ...
## $ Gender.Inequality.Index..2018.: num 0.671 0.537 0.164 NA 0.103 0.315 0.26 NA 0.092 0.061 ...
## $ Gender.Inequality.Index..2019.: num 0.665 0.537 0.156 NA 0.056 0.306 0.239 NA 0.08 0.054 ...
## $ Gender.Inequality.Index..2020.: num 0.674 0.537 0.156 NA 0.05 0.293 0.239 NA 0.074 0.052 ...
## $ Gender.Inequality.Index..2021.: num 0.678 0.537 0.144 NA 0.049 0.287 0.216 NA 0.073 0.053 ...
```

```
summary(data)
```

```

##      ISO3          Country          Continent          Hemisphere
## Length:195      Length:195      Length:195      Length:195
## Class :character Class :character Class :character Class :character
## Mode  :character Mode  :character Mode  :character Mode  :character
##
##
##
##
## Human.Development.Groups UNDP.Developing.Regions HDI.Rank..2021.
## Length:195      Length:195      Min.   : 1.00
## Class :character Class :character 1st Qu.: 48.50
## Mode  :character Mode  :character Median : 96.00
##                                     Mean  : 95.81
##                                     3rd Qu.:143.50
##                                     Max.   :191.00
##                                     NA's    :4
##
## GII.Rank..2021.  Gender.Inequality.Index..1990. Gender.Inequality.Index..1991.
## Min.   : 1.00   Min.   :0.0900   Min.   :0.0900
## 1st Qu.: 43.25   1st Qu.:0.3195   1st Qu.:0.3277
## Median : 85.50   Median :0.5150   Median :0.5075
## Mean   : 85.38   Mean   :0.4876   Mean   :0.4851
## 3rd Qu.:127.75   3rd Qu.:0.6590   3rd Qu.:0.6495
## Max.   :170.00   Max.   :0.8110   Max.   :0.8060
## NA's    :25     NA's    :67     NA's    :67
##
## Gender.Inequality.Index..1992. Gender.Inequality.Index..1993.
## Min.   :0.0940   Min.   :0.0900
## 1st Qu.:0.3292   1st Qu.:0.3250
## Median :0.5090   Median :0.5070
## Mean   :0.4818   Mean   :0.4785
## 3rd Qu.:0.6448   3rd Qu.:0.6438
## Max.   :0.8030   Max.   :0.8020
## NA's    :67     NA's    :67
##
## Gender.Inequality.Index..1994. Gender.Inequality.Index..1995.
## Min.   :0.0830   Min.   :0.0750
## 1st Qu.:0.3172   1st Qu.:0.3145
## Median :0.5000   Median :0.4925
## Mean   :0.4747   Mean   :0.4706
## 3rd Qu.:0.6385   3rd Qu.:0.6330
## Max.   :0.8000   Max.   :0.7990
## NA's    :67     NA's    :65
##
## Gender.Inequality.Index..1996. Gender.Inequality.Index..1997.
## Min.   :0.0670   Min.   :0.0630
## 1st Qu.:0.3008   1st Qu.:0.2853
## Median :0.4855   Median :0.4795
## Mean   :0.4656   Mean   :0.4605
## 3rd Qu.:0.6278   3rd Qu.:0.6180
## Max.   :0.7980   Max.   :0.7960
## NA's    :65     NA's    :65
##
## Gender.Inequality.Index..1998. Gender.Inequality.Index..1999.
## Min.   :0.0600   Min.   :0.0590
## 1st Qu.:0.2980   1st Qu.:0.2840
## Median :0.4805   Median :0.4820
## Mean   :0.4594   Mean   :0.4548
## 3rd Qu.:0.6178   3rd Qu.:0.6240
## Max.   :0.8120   Max.   :0.8070
## NA's    :59     NA's    :54
##
## Gender.Inequality.Index..2000. Gender.Inequality.Index..2001.

```

## Min. :0.0600	Min. :0.0560
## 1st Qu.:0.2780	1st Qu.:0.2710
## Median :0.4615	Median :0.4640
## Mean :0.4487	Mean :0.4443
## 3rd Qu.:0.6162	3rd Qu.:0.6120
## Max. :0.8050	Max. :0.8030
## NA's :51	NA's :46
## Gender.Inequality.Index..2002.	Gender.Inequality.Index..2003.
## Min. :0.0540	Min. :0.0490
## 1st Qu.:0.2625	1st Qu.:0.2610
## Median :0.4665	Median :0.4640
## Mean :0.4387	Mean :0.4338
## 3rd Qu.:0.6050	3rd Qu.:0.5970
## Max. :0.8020	Max. :0.8220
## NA's :45	NA's :42
## Gender.Inequality.Index..2004.	Gender.Inequality.Index..2005.
## Min. :0.0480	Min. :0.0470
## 1st Qu.:0.2567	1st Qu.:0.2580
## Median :0.4670	Median :0.4540
## Mean :0.4284	Mean :0.4232
## 3rd Qu.:0.5960	3rd Qu.:0.5930
## Max. :0.8210	Max. :0.7990
## NA's :39	NA's :36
## Gender.Inequality.Index..2006.	Gender.Inequality.Index..2007.
## Min. :0.0470	Min. :0.0470
## 1st Qu.:0.2487	1st Qu.:0.2440
## Median :0.4340	Median :0.4220
## Mean :0.4144	Mean :0.4062
## 3rd Qu.:0.5880	3rd Qu.:0.5720
## Max. :0.8000	Max. :0.7990
## NA's :33	NA's :32
## Gender.Inequality.Index..2008.	Gender.Inequality.Index..2009.
## Min. :0.0480	Min. :0.0490
## 1st Qu.:0.2440	1st Qu.:0.2320
## Median :0.4240	Median :0.4280
## Mean :0.4039	Mean :0.4000
## 3rd Qu.:0.5650	3rd Qu.:0.5595
## Max. :0.7980	Max. :0.7980
## NA's :32	NA's :32
## Gender.Inequality.Index..2010.	Gender.Inequality.Index..2011.
## Min. :0.0490	Min. :0.0440
## 1st Qu.:0.2310	1st Qu.:0.2270
## Median :0.4220	Median :0.4240
## Mean :0.3987	Mean :0.3939
## 3rd Qu.:0.5580	3rd Qu.:0.5555
## Max. :0.7980	Max. :0.7990
## NA's :30	NA's :28
## Gender.Inequality.Index..2012.	Gender.Inequality.Index..2013.
## Min. :0.0420	Min. :0.0390
## 1st Qu.:0.2290	1st Qu.:0.2165
## Median :0.4090	Median :0.3980
## Mean :0.3875	Mean :0.3797
## 3rd Qu.:0.5515	3rd Qu.:0.5450
## Max. :0.8000	Max. :0.8020
## NA's :28	NA's :28
## Gender.Inequality.Index..2014.	Gender.Inequality.Index..2015.
## Min. :0.0350	Min. :0.0330
## 1st Qu.:0.2180	1st Qu.:0.2115

```
## Median :0.3960      Median :0.3950
## Mean   :0.3772      Mean    :0.3718
## 3rd Qu.:0.5450      3rd Qu.:0.5327
## Max.   :0.8060      Max.    :0.8150
## NA's   :28          NA's    :27
## Gender.Inequality.Index..2016. Gender.Inequality.Index..2017.
## Min.   :0.0300      Min.    :0.0230
## 1st Qu.:0.1920      1st Qu.:0.1893
## Median :0.3900      Median :0.3780
## Mean   :0.3652      Mean    :0.3559
## 3rd Qu.:0.5270      3rd Qu.:0.5165
## Max.   :0.8100      Max.    :0.8080
## NA's   :26          NA's    :25
## Gender.Inequality.Index..2018. Gender.Inequality.Index..2019.
## Min.   :0.0180      Min.    :0.0160
## 1st Qu.:0.1832      1st Qu.:0.1710
## Median :0.3785      Median :0.3700
## Mean   :0.3518      Mean    :0.3469
## 3rd Qu.:0.5162      3rd Qu.:0.5105
## Max.   :0.8060      Max.    :0.7850
## NA's   :25          NA's    :25
## Gender.Inequality.Index..2020. Gender.Inequality.Index..2021.
## Min.   :0.0130      Min.    :0.0130
## 1st Qu.:0.1713      1st Qu.:0.1775
## Median :0.3660      Median :0.3630
## Mean   :0.3442      Mean    :0.3444
## 3rd Qu.:0.5035      3rd Qu.:0.5058
## Max.   :0.7840      Max.    :0.8200
## NA's   :25          NA's    :25
```

```
# Load necessary Library
#install.packages("ggplot2")
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 4.4.2
```

```
# Filter data for a specific country
afghanistan_data <- data[data$Country == "India", ]

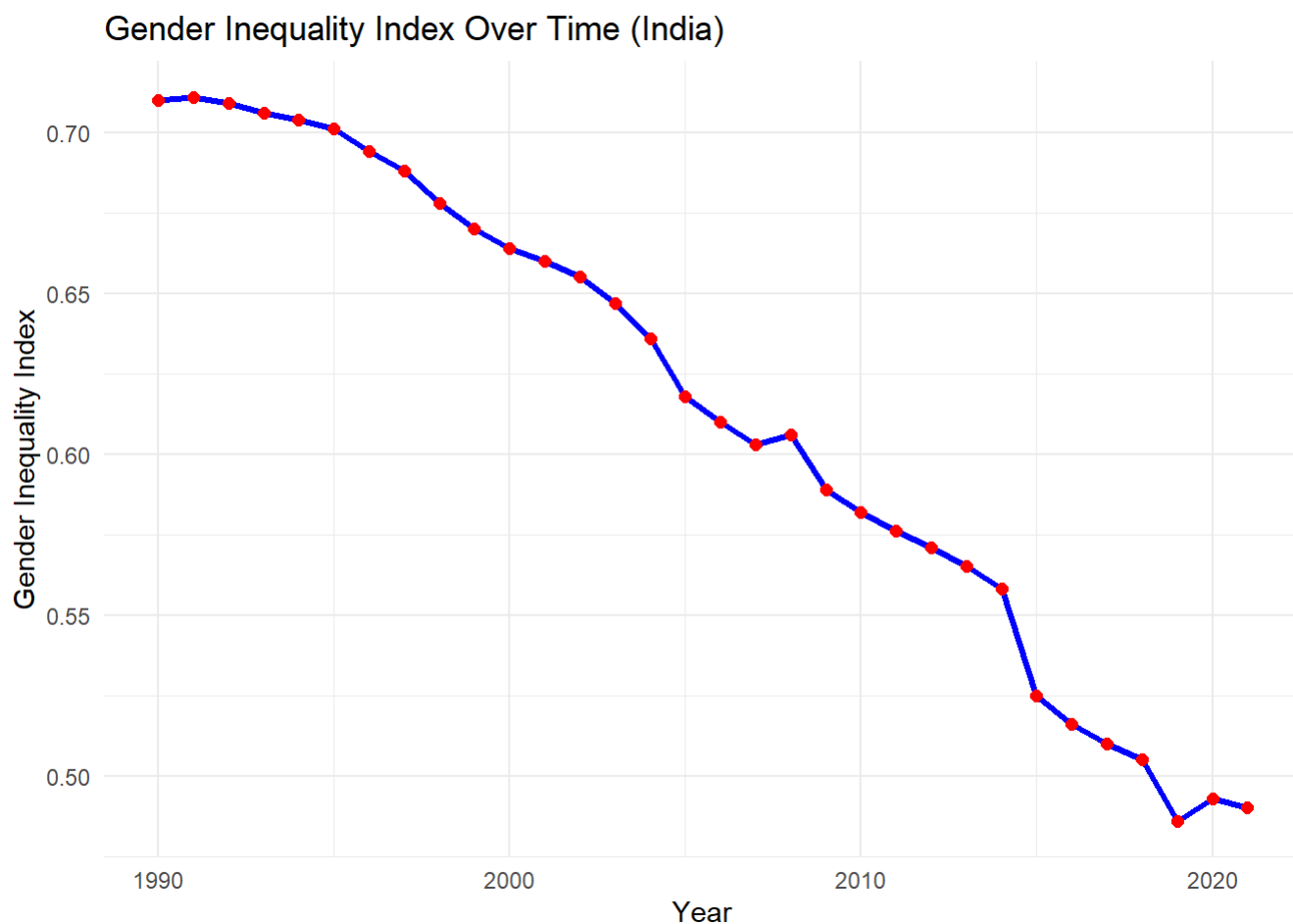
# Create a Line plot for Gender Inequality Index over time
years <- 1990:2021
gii_values <- as.numeric(afghanistan_data[, 9:40]) # Select GII columns

# Create a data frame for plotting
plot_data <- data.frame(Year = years, GII = gii_values)

# Plot the data
ggplot(plot_data, aes(x = Year, y = GII)) +
  geom_line(color = "blue", size = 1.2) +
  geom_point(color = "red", size = 2) +
  labs(title = "Gender Inequality Index Over Time (India)",
       x = "Year", y = "Gender Inequality Index") +
  theme_minimal()
```

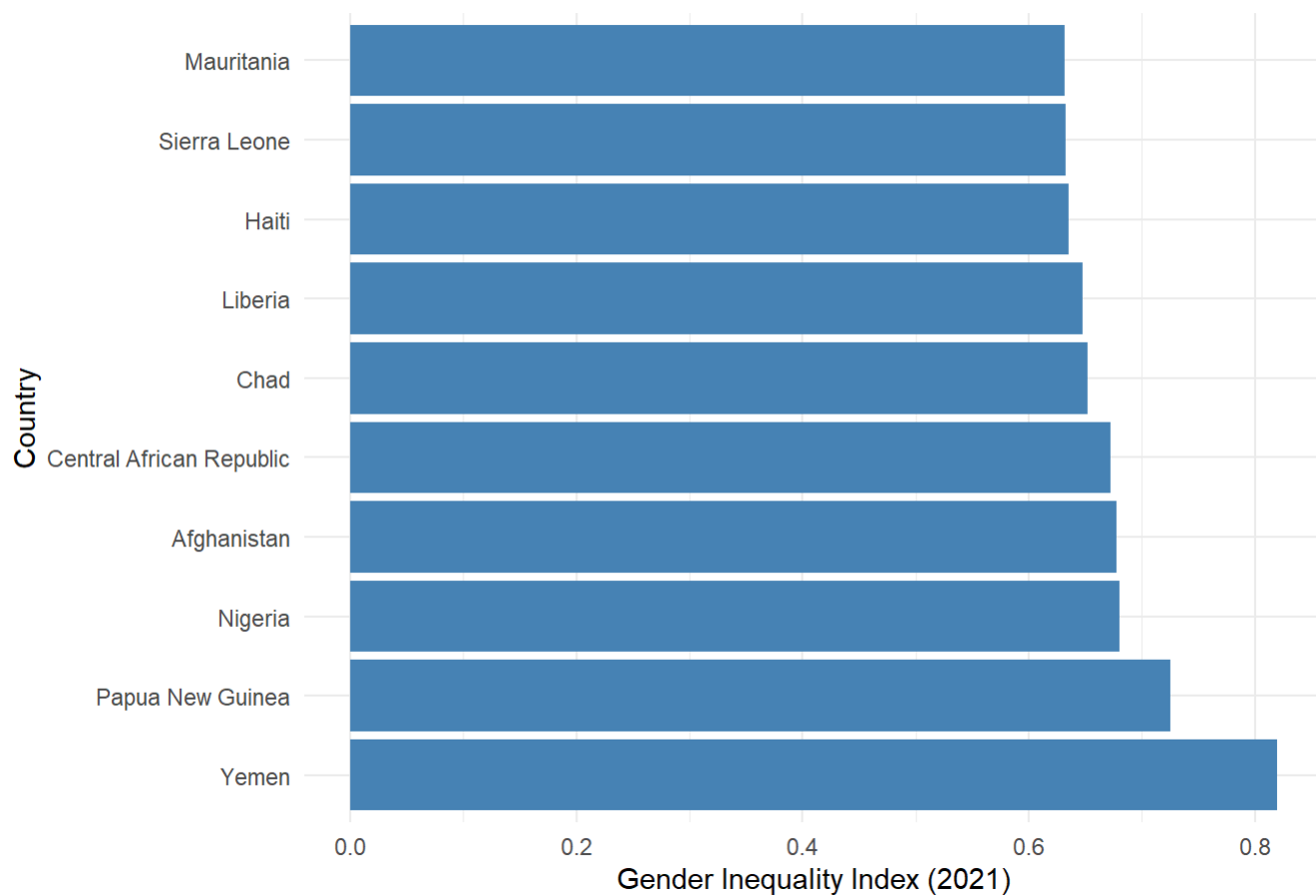


```
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.  
## i Please use `linewidth` instead.  
## This warning is displayed once every 8 hours.  
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was  
## generated.
```



```
library(ggplot2)  
  
# Select relevant columns: Country and GII for 2021  
gii_2021 <- data[, c("Country", "Gender.Inequality.Index..2021.")]  
  
# Remove missing values  
gii_2021 <- na.omit(gii_2021)  
  
# Create a bar plot for GII in 2021 for a subset of countries (e.g., top 10)  
top_countries <- head(gii_2021[order(gii_2021$Gender.Inequality.Index..2021., decreasing = TRUE), ], 10)  
  
# Plot the data  
ggplot(top_countries, aes(x = reorder(Country, -Gender.Inequality.Index..2021.), y = Gender.Inequality.Index..2021.)) +  
  geom_bar(stat = "identity", fill = "steelblue") +  
  labs(title = "Top 10 Countries with Highest Gender Inequality Index (2021)",  
       x = "Country", y = "Gender Inequality Index (2021)") +  
  coord_flip() +  
  theme_minimal()
```

Top 10 Countries with Highest Gender Inequality Index (2021)



```
##ARIMA MODEL
library(forecast)
```

```
## Warning: package 'forecast' was built under R version 4.4.2
```

```
## Registered S3 method overwritten by 'quantmod':
##   method      from
##   as.zoo.data.frame zoo
```

```
# Adjusted grep pattern to match the format "Gender.Inequality.Index..YYYY."
gii_columns <- grep("^Gender\\.Inequality\\.Index\\.\\.\\.\\.\\d{4}\\.$", names(data), value = TRUE)
print(gii_columns) # Print the columns found to confirm
```

```
## [1] "Gender.Inequality.Index..1990." "Gender.Inequality.Index..1991."
## [3] "Gender.Inequality.Index..1992." "Gender.Inequality.Index..1993."
## [5] "Gender.Inequality.Index..1994." "Gender.Inequality.Index..1995."
## [7] "Gender.Inequality.Index..1996." "Gender.Inequality.Index..1997."
## [9] "Gender.Inequality.Index..1998." "Gender.Inequality.Index..1999."
## [11] "Gender.Inequality.Index..2000." "Gender.Inequality.Index..2001."
## [13] "Gender.Inequality.Index..2002." "Gender.Inequality.Index..2003."
## [15] "Gender.Inequality.Index..2004." "Gender.Inequality.Index..2005."
## [17] "Gender.Inequality.Index..2006." "Gender.Inequality.Index..2007."
## [19] "Gender.Inequality.Index..2008." "Gender.Inequality.Index..2009."
## [21] "Gender.Inequality.Index..2010." "Gender.Inequality.Index..2011."
## [23] "Gender.Inequality.Index..2012." "Gender.Inequality.Index..2013."
## [25] "Gender.Inequality.Index..2014." "Gender.Inequality.Index..2015."
## [27] "Gender.Inequality.Index..2016." "Gender.Inequality.Index..2017."
## [29] "Gender.Inequality.Index..2018." "Gender.Inequality.Index..2019."
## [31] "Gender.Inequality.Index..2020." "Gender.Inequality.Index..2021."
```

```
# Check if any GII columns were found
if (length(gii_columns) == 0) {
  stop("No GII columns found. Please check the column names in the dataset.")
}

# Calculate the average GII for each year across all countries
gii_yearly_avg <- colMeans(data[, gii_columns], na.rm = TRUE)
print(gii_yearly_avg) # Print the yearly averages to ensure they contain values
```

```
## Gender.Inequality.Index..1990. Gender.Inequality.Index..1991.
##          0.4875781          0.4850781
## Gender.Inequality.Index..1992. Gender.Inequality.Index..1993.
##          0.4817656          0.4784844
## Gender.Inequality.Index..1994. Gender.Inequality.Index..1995.
##          0.4747422          0.4706154
## Gender.Inequality.Index..1996. Gender.Inequality.Index..1997.
##          0.4656154          0.4605154
## Gender.Inequality.Index..1998. Gender.Inequality.Index..1999.
##          0.4593750          0.4548085
## Gender.Inequality.Index..2000. Gender.Inequality.Index..2001.
##          0.4487153          0.4442685
## Gender.Inequality.Index..2002. Gender.Inequality.Index..2003.
##          0.4387133          0.4337908
## Gender.Inequality.Index..2004. Gender.Inequality.Index..2005.
##          0.4284103          0.4232390
## Gender.Inequality.Index..2006. Gender.Inequality.Index..2007.
##          0.4144383          0.4062086
## Gender.Inequality.Index..2008. Gender.Inequality.Index..2009.
##          0.4039325          0.3999939
## Gender.Inequality.Index..2010. Gender.Inequality.Index..2011.
##          0.3987152          0.3939102
## Gender.Inequality.Index..2012. Gender.Inequality.Index..2013.
##          0.3875030          0.3797186
## Gender.Inequality.Index..2014. Gender.Inequality.Index..2015.
##          0.3772275          0.3717857
## Gender.Inequality.Index..2016. Gender.Inequality.Index..2017.
##          0.3652130          0.3558882
## Gender.Inequality.Index..2018. Gender.Inequality.Index..2019.
##          0.3517824          0.3469176
## Gender.Inequality.Index..2020. Gender.Inequality.Index..2021.
##          0.3441588          0.3443765
```

```
# Ensure there is data to proceed
if (length(gii_yearly_avg) == 0) {
  stop("No GII data found for the selected columns.")
}

# Convert the averages to a time series (starting in 1990, with yearly frequency)
gii_ts <- ts(gii_yearly_avg, start = 1990, frequency = 1)

# Fit an ARIMA model
model <- auto.arima(gii_ts)

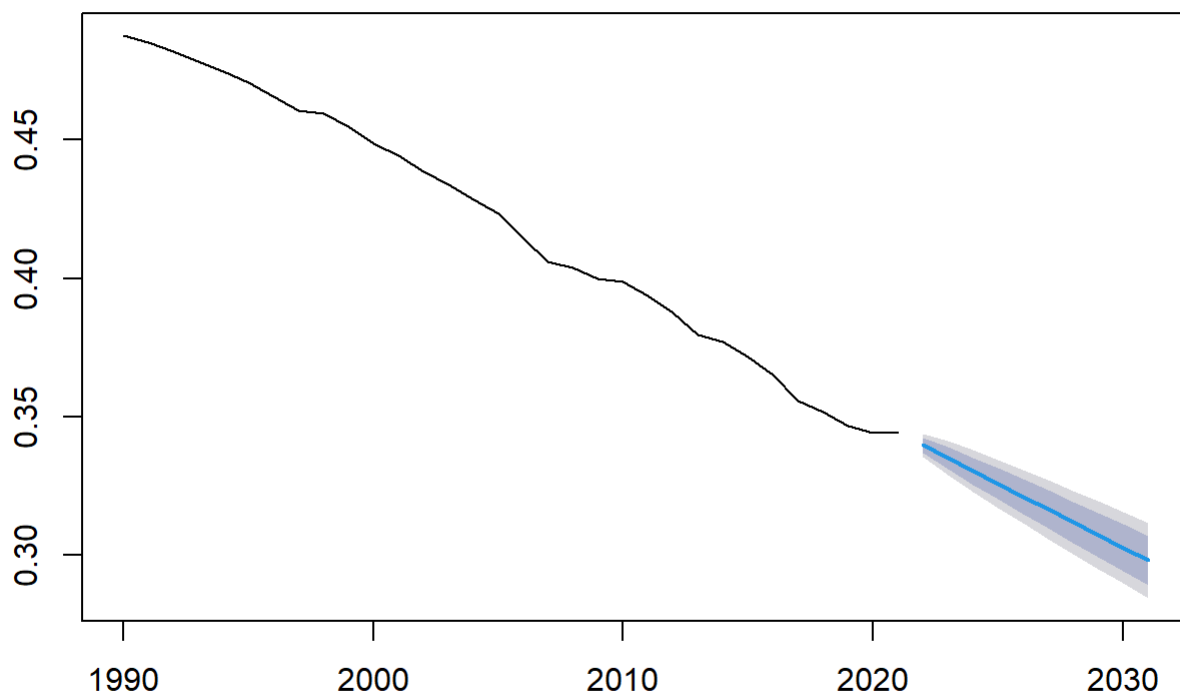
# Forecast for the next 10 years
forecast_values <- forecast(model, h = 10)
forecast_values
```

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 2022	0.3397571	0.3369395	0.3425746	0.3354480	0.3440661
## 2023	0.3351377	0.3311531	0.3391222	0.3290438	0.3412316
## 2024	0.3305182	0.3256381	0.3353983	0.3230548	0.3379817
## 2025	0.3258988	0.3202638	0.3315339	0.3172808	0.3345169
## 2026	0.3212794	0.3149792	0.3275796	0.3116441	0.3309147
## 2027	0.3166600	0.3097585	0.3235615	0.3061051	0.3272150
## 2028	0.3120406	0.3045861	0.3194951	0.3006400	0.3234413
## 2029	0.3074212	0.2994520	0.3153904	0.2952334	0.3196090
## 2030	0.3028018	0.2943492	0.3112544	0.2898747	0.3157289
## 2031	0.2981824	0.2892726	0.3070922	0.2845560	0.3118088

```
# Extract the predicted values
predicted_values <- as.numeric(forecast_values$mean)

# Plot the forecast
plot(forecast_values, main = "Arima Model Forecast of Global Average GII for the Next 10 Years")
```

Arima Model Forecast of Global Average GII for the Next 10 Years



```
# Create a vector of actual GII values for the next 10 years (replace with real values if available)
actual_future_gii <- c(0.339, 0.335, 0.3305, 0.32, 0.321, 0.316, 0.312, 0.307, 0.302, 0.298)

# Create a data frame for evaluation
future_years <- data.frame(Year = (2022:2031))
gii_evaluation <- data.frame(
  Year = future_years$Year,
  Predicted = predicted_values,
  Actual = actual_future_gii
)

# Print the evaluation table
print(gii_evaluation)
```

```
##      Year Predicted Actual
## 1  2022 0.3397571 0.3390
## 2  2023 0.3351377 0.3350
## 3  2024 0.3305182 0.3305
## 4  2025 0.3258988 0.3200
## 5  2026 0.3212794 0.3210
## 6  2027 0.3166600 0.3160
## 7  2028 0.3120406 0.3120
## 8  2029 0.3074212 0.3070
## 9  2030 0.3028018 0.3020
## 10 2031 0.2981824 0.2980
```

```
# Calculate performance metrics
MAE <- mean(abs(gii_evaluation$Actual - gii_evaluation$Predicted))
MSE <- mean((gii_evaluation$Actual - gii_evaluation$Predicted)^2)
RMSE <- sqrt(MSE)
R_squared <- 1 - (sum((gii_evaluation$Actual - gii_evaluation$Predicted)^2) /
  sum((gii_evaluation$Actual - mean(gii_evaluation$Actual))^2))

# Print the performance metrics
print("ARIMA Model Performance Metrics")
```

```
## [1] "ARIMA Model Performance Metrics"
```

```
print(paste("Mean Absolute Error (MAE):", round(MAE, 4)))
```

```
## [1] "Mean Absolute Error (MAE): 9e-04"
```

```
print(paste("Mean Squared Error (MSE):", round(MSE, 4)))
```

```
## [1] "Mean Squared Error (MSE): 0"
```

```
print(paste("Root Mean Squared Error (RMSE):", round(RMSE, 4)))
```

```
## [1] "Root Mean Squared Error (RMSE): 0.0019"
```

```
print(paste("R-squared:", round(R_squared, 4)))
```

```
## [1] "R-squared: 0.9786"
```

```
#LINEAR REGRESSION
```

```
# Convert the years to a numeric vector and create a data frame for modeling
```

```
years <- 1990:(1990 + length(gii_yearly_avg) - 1)
```

```
gii_df <- data.frame(Year = years, GII = gii_yearly_avg)
```

```
# Fit a linear regression model
```

```
linear_model <- lm(GII ~ Year, data = gii_df)
```

```
# Predict the next 10 years
```

```
future_years <- data.frame(Year = (max(years) + 1):(max(years) + 10))
```

```
predictions <- predict(linear_model, newdata = future_years)
```

```
predictions
```

```
##           1           2           3           4           5           6           7           8
## 0.3362163 0.3312569 0.3262975 0.3213381 0.3163787 0.3114193 0.3064599 0.3015005
##           9          10
## 0.2965411 0.2915817
```

```
# Combine past data with predictions for plotting
```

```
gii_combined <- c(gii_yearly_avg, predictions)
```

```
years_combined <- c(years, future_years$Year)
```

```
# Plot the historical data and predictions
```

```
plot(years_combined, gii_combined, type = "l", col = "blue",
```

```
      xlab = "Year", ylab = "Gender Inequality Index",
```

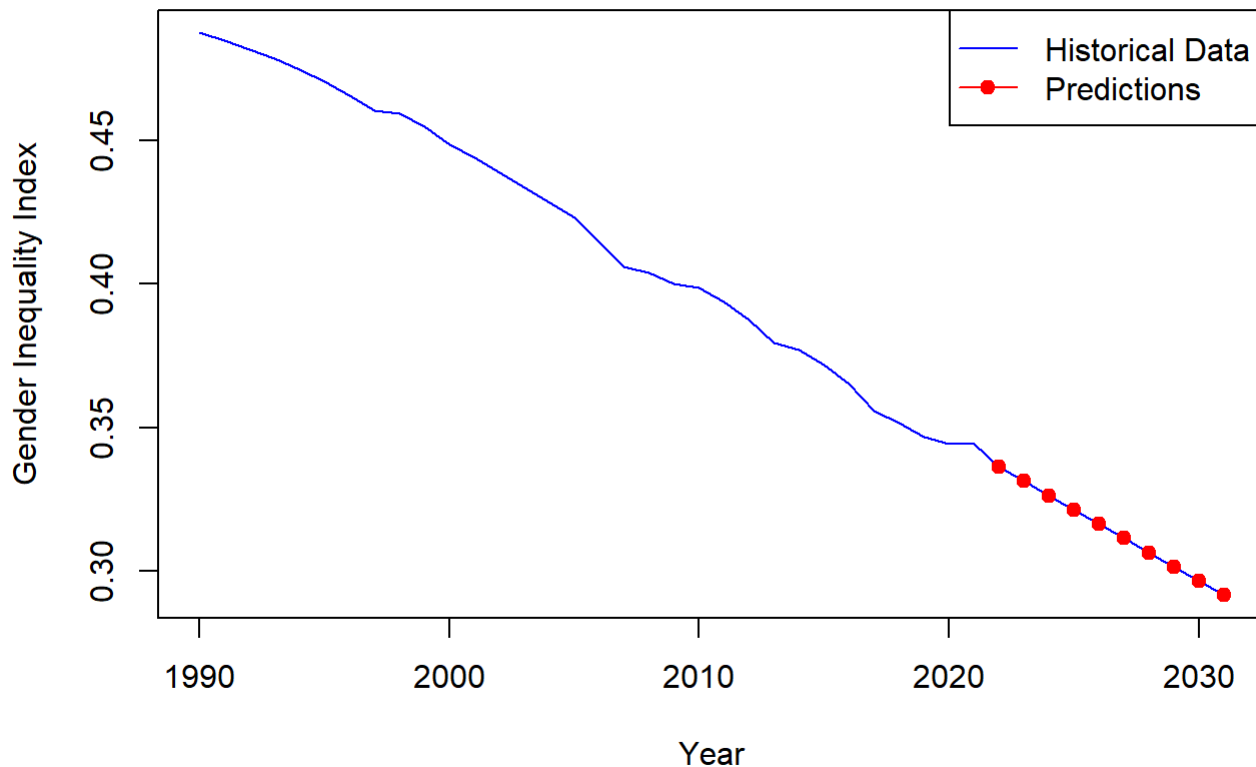
```
      main = "Linear Regression Forecast of Global Average GII for the Next 10 Years")
```

```
points(future_years$Year, predictions, col = "red", pch = 19)
```

```
legend("topright", legend = c("Historical Data", "Predictions"),
```

```
      col = c("blue", "red"), lty = 1, pch = c(NA, 19))
```

Linear Regression Forecast of Global Average GII for the Next 10 Years



```
# Assuming you have actual GII data for the years you predicted
# Create a vector of actual GII values for the next 4 years (for example purposes)
actual_future_gii <- c(0.336,0.331,0.326,0.3213,0.316,0.311,0.306,0.301,0.296,0.291) # Replace with your actual values

# Combine actual values with predicted values for evaluation
gii_evaluation <- data.frame(Year = future_years$Year,
                             Predicted = predictions,
                             Actual = actual_future_gii)

# Calculate performance metrics
MAE <- mean(abs(gii_evaluation$Actual - gii_evaluation$Predicted))
MSE <- mean((gii_evaluation$Actual - gii_evaluation$Predicted)^2)
RMSE <- sqrt(MSE)
R_squared <- 1 - (sum((gii_evaluation$Actual - gii_evaluation$Predicted)^2) /
                 sum((gii_evaluation$Actual - mean(gii_evaluation$Actual))^2))

# Print the performance metrics
print(paste("Linear Regression Model"))
```

```
## [1] "Linear Regression Model"
```

```
print(paste("Mean Absolute Error (MAE):", MAE))
```

```
## [1] "Mean Absolute Error (MAE): 0.00036898965770229"
```

```
print(paste("Mean Squared Error (MSE):", MSE))
```



```
## [1] "Mean Squared Error (MSE): 1.61505505738178e-07"
```

```
print(paste("Root Mean Squared Error (RMSE):", RMSE))
```

```
## [1] "Root Mean Squared Error (RMSE): 0.000401877476027431"
```

```
print(paste("R-squared:", R_squared))
```

```
## [1] "R-squared: 0.999218678388809"
```

```
#ENSEMBLE MODEL
```

```
# Adjust grep pattern to match column format "Gender.Inequality.Index..YYYY."
```

```
gii_columns <- grep("^Gender\\.\\.Inequality\\.\\.Index\\.\\.\\.\\.\\d{4}\\.$", names(data), value = TRUE)
```

```
print(gii_columns)
```

```
## [1] "Gender.Inequality.Index..1990." "Gender.Inequality.Index..1991."  
## [3] "Gender.Inequality.Index..1992." "Gender.Inequality.Index..1993."  
## [5] "Gender.Inequality.Index..1994." "Gender.Inequality.Index..1995."  
## [7] "Gender.Inequality.Index..1996." "Gender.Inequality.Index..1997."  
## [9] "Gender.Inequality.Index..1998." "Gender.Inequality.Index..1999."  
## [11] "Gender.Inequality.Index..2000." "Gender.Inequality.Index..2001."  
## [13] "Gender.Inequality.Index..2002." "Gender.Inequality.Index..2003."  
## [15] "Gender.Inequality.Index..2004." "Gender.Inequality.Index..2005."  
## [17] "Gender.Inequality.Index..2006." "Gender.Inequality.Index..2007."  
## [19] "Gender.Inequality.Index..2008." "Gender.Inequality.Index..2009."  
## [21] "Gender.Inequality.Index..2010." "Gender.Inequality.Index..2011."  
## [23] "Gender.Inequality.Index..2012." "Gender.Inequality.Index..2013."  
## [25] "Gender.Inequality.Index..2014." "Gender.Inequality.Index..2015."  
## [27] "Gender.Inequality.Index..2016." "Gender.Inequality.Index..2017."  
## [29] "Gender.Inequality.Index..2018." "Gender.Inequality.Index..2019."  
## [31] "Gender.Inequality.Index..2020." "Gender.Inequality.Index..2021."
```

```
# Check if GII columns were found
```

```
if (length(gii_columns) == 0) {
```

```
  stop("No GII columns found. Please check the column names in the dataset.")
```

```
}
```

```
# Calculate the average GII for each year across all countries
```

```
gii_yearly_avg <- colMeans(data[, gii_columns], na.rm = TRUE)
```

```
print(gii_yearly_avg)
```

## Gender.Inequality.Index..1990.	Gender.Inequality.Index..1991.
## 0.4875781	0.4850781
## Gender.Inequality.Index..1992.	Gender.Inequality.Index..1993.
## 0.4817656	0.4784844
## Gender.Inequality.Index..1994.	Gender.Inequality.Index..1995.
## 0.4747422	0.4706154
## Gender.Inequality.Index..1996.	Gender.Inequality.Index..1997.
## 0.4656154	0.4605154
## Gender.Inequality.Index..1998.	Gender.Inequality.Index..1999.
## 0.4593750	0.4548085
## Gender.Inequality.Index..2000.	Gender.Inequality.Index..2001.
## 0.4487153	0.4442685
## Gender.Inequality.Index..2002.	Gender.Inequality.Index..2003.
## 0.4387133	0.4337908
## Gender.Inequality.Index..2004.	Gender.Inequality.Index..2005.
## 0.4284103	0.4232390
## Gender.Inequality.Index..2006.	Gender.Inequality.Index..2007.
## 0.4144383	0.4062086
## Gender.Inequality.Index..2008.	Gender.Inequality.Index..2009.
## 0.4039325	0.3999939
## Gender.Inequality.Index..2010.	Gender.Inequality.Index..2011.
## 0.3987152	0.3939102
## Gender.Inequality.Index..2012.	Gender.Inequality.Index..2013.
## 0.3875030	0.3797186
## Gender.Inequality.Index..2014.	Gender.Inequality.Index..2015.
## 0.3772275	0.3717857
## Gender.Inequality.Index..2016.	Gender.Inequality.Index..2017.
## 0.3652130	0.3558882
## Gender.Inequality.Index..2018.	Gender.Inequality.Index..2019.
## 0.3517824	0.3469176
## Gender.Inequality.Index..2020.	Gender.Inequality.Index..2021.
## 0.3441588	0.3443765

```

# Ensure data is valid
if (length(gii_yearly_avg) == 0) {
  stop("No GII data found for the selected columns.")
}

# Convert to a time series object
gii_ts <- ts(gii_yearly_avg, start = 1990, frequency = 1)

### ARIMA Model
arima_model <- auto.arima(gii_ts)
arima_forecast <- forecast(arima_model, h = 10)
arima_predictions <- as.numeric(arima_forecast$mean)

### Exponential Smoothing Model
ets_model <- ets(gii_ts)
ets_forecast <- forecast(ets_model, h = 10)
ets_predictions <- as.numeric(ets_forecast$mean)

### Linear Regression Model
years <- as.numeric(time(gii_ts))
lm_model <- lm(gii_yearly_avg ~ years)
future_years <- data.frame(years = seq(max(years) + 1, max(years) + 10))
lm_predictions <- predict(lm_model, newdata = future_years)

### Ensemble Model
# Take the average of predictions from the three models
ensemble_predictions <- rowMeans(cbind(arima_predictions, ets_predictions, lm_predictions))

# Create a data frame for all predictions
forecast_df <- data.frame(
  Year = seq(2022, 2031),
  ARIMA = arima_predictions,
  ETS = ets_predictions,
  Linear_Regression = lm_predictions,
  Ensemble = ensemble_predictions
)

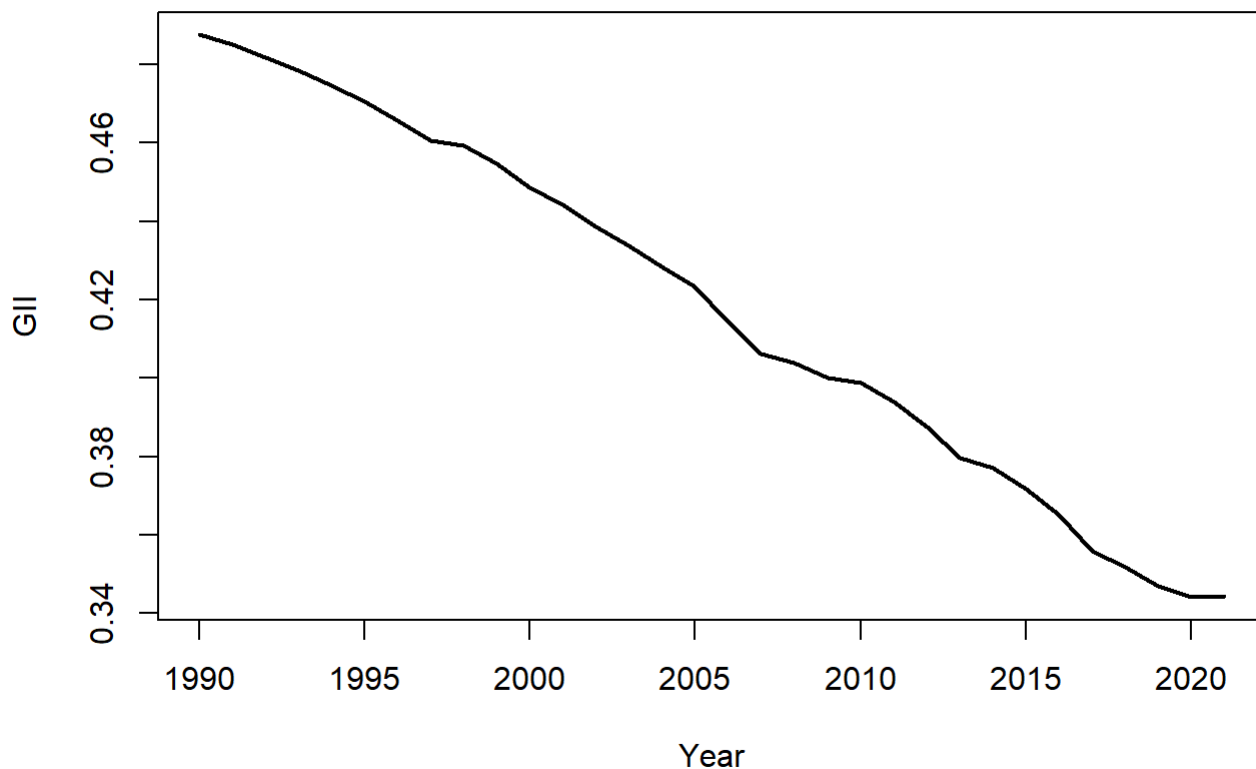
print(forecast_df)

```

##	Year	ARIMA	ETS	Linear_Regression	Ensemble
## 1	2022	0.3397571	0.339759	0.3362163	0.3385774
## 2	2023	0.3351377	0.335142	0.3312569	0.3338455
## 3	2024	0.3305182	0.330525	0.3262975	0.3291136
## 4	2025	0.3258988	0.325908	0.3213381	0.3243816
## 5	2026	0.3212794	0.321291	0.3163787	0.3196497
## 6	2027	0.3166600	0.316674	0.3114193	0.3149178
## 7	2028	0.3120406	0.312057	0.3064599	0.3101858
## 8	2029	0.3074212	0.307440	0.3015005	0.3054539
## 9	2030	0.3028018	0.302823	0.2965411	0.3007220
## 10	2031	0.2981824	0.298206	0.2915817	0.2959900

```
# Plot the predictions
plot(
  seq(1990, 2021), gii_yearly_avg, type = "l", col = "black", lwd = 2,
  xlab = "Year", ylab = "GII", main = "GII Forecasts Using Ensemble Model"
)
```

GII Forecasts Using Ensemble Model



```
### Evaluate Performance (Optional: Use actual future GII values if available)
actual_future_gii <- c(0.339, 0.335, 0.3305, 0.32, 0.321, 0.316, 0.312, 0.307, 0.302, 0.298) # Replace
with real data if available
evaluation_df <- data.frame(
  Year = forecast_df$Year,
  Predicted_Ensemble = ensemble_predictions,
  Actual = actual_future_gii
)

# Calculate performance metrics
MAE <- mean(abs(evaluation_df$Actual - evaluation_df$Predicted_Ensemble))
MSE <- mean((evaluation_df$Actual - evaluation_df$Predicted_Ensemble)^2)
RMSE <- sqrt(MSE)
R_squared <- 1 - (sum((evaluation_df$Actual - evaluation_df$Predicted_Ensemble)^2) /
  sum((evaluation_df$Actual - mean(evaluation_df$Actual))^2))

# Print performance metrics
print("Ensemble Model Performance Metrics")
```

```
## [1] "Ensemble Model Performance Metrics"
```

```
print(paste("Mean Absolute Error (MAE):", round(MAE, 4)))
```

```
## [1] "Mean Absolute Error (MAE): 0.0016"
```

```
print(paste("Mean Squared Error (MSE):", round(MSE, 4)))
```

```
## [1] "Mean Squared Error (MSE): 0"
```

```
print(paste("Root Mean Squared Error (RMSE):", round(RMSE, 4)))
```

```
## [1] "Root Mean Squared Error (RMSE): 0.0019"
```

```
print(paste("R-squared:", round(R_squared, 4)))
```

```
## [1] "R-squared: 0.9785"
```

```
#NEURAL NETWORK  
#install.packages("nnet")  
library(nnet)
```

```
## Warning: package 'nnet' was built under R version 4.4.2
```

```
# Adjust grep pattern to match columns for GII values (e.g., "Gender.Inequality.Index..1990.")  
gii_columns <- grep("^Gender\\.Inequality\\.Index\\.\\.\\.\\.\\d{4}\\.$", names(data), value = TRUE)
```

```
# Check if columns are correctly identified  
if (length(gii_columns) == 0) {  
  stop("No GII columns found. Please check the column names in the dataset.")  
}
```

```
# Calculate the average GII for each year across all countries  
gii_yearly_avg <- colMeans(data[, gii_columns], na.rm = TRUE)
```

```
# Ensure there is data to proceed  
if (length(gii_yearly_avg) == 0) {  
  stop("No data available for GII.")  
}
```

```
# Prepare the data  
years <- as.numeric(sub("Gender\\.Inequality\\.Index\\.\\.\\.\\.\\d{4}\\.$", "\\1", gii_columns))  
gii_data <- data.frame(Year = years, GII = gii_yearly_avg)
```

```
# Normalize the data for neural network training  
normalize <- function(x) (x - min(x)) / (max(x) - min(x))  
gii_data$GII_norm <- normalize(gii_data$GII)
```

```
# Train/Test Split  
train_data <- gii_data[gii_data$Year <= max(gii_data$Year) - 10, ]  
test_data <- gii_data[gii_data$Year > max(gii_data$Year) - 10, ]
```

```
# Neural Network Model  
set.seed(123) # For reproducibility  
nn_model <- nnet(GII_norm ~ Year, data = train_data, size = 3, linout = TRUE, skip = TRUE)
```

```
## # weights: 11
## initial value 36012684.715082
## final value 0.009366
## converged
```

```
# Predict on the training data
train_data$Predicted_norm <- predict(nn_model, train_data)

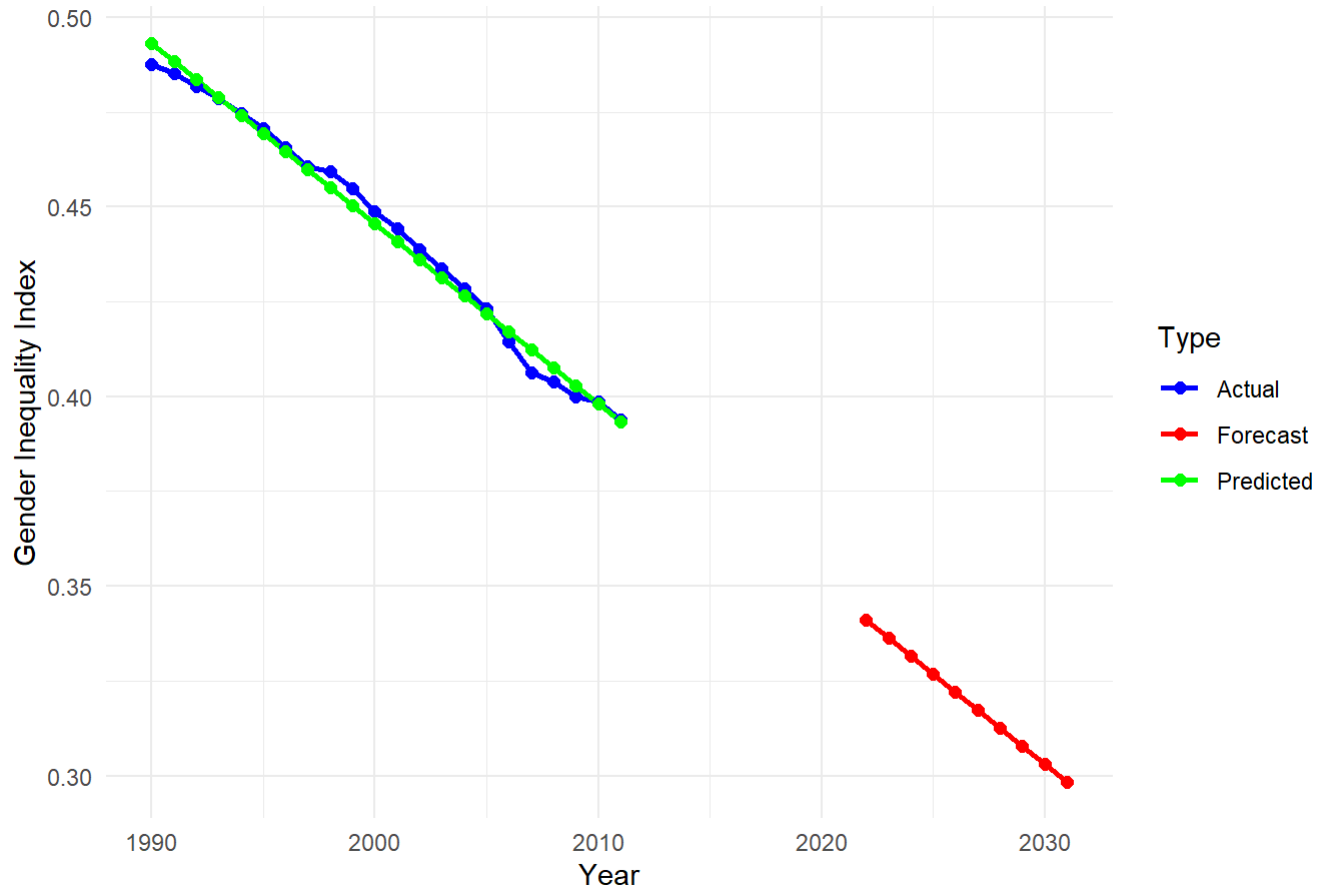
# De-normalize predictions
denormalize <- function(x, orig) x * (max(orig) - min(orig)) + min(orig)
train_data$Predicted <- denormalize(train_data$Predicted_norm, gii_data$GII)

# Predict the next 10 years
future_years <- data.frame(Year = seq(max(gii_data$Year) + 1, max(gii_data$Year) + 10))
future_years$Predicted_norm <- predict(nn_model, future_years)
future_years$Predicted <- denormalize(future_years$Predicted_norm, gii_data$GII)

# Combine train, test, and future data for visualization
combined_data <- rbind(
  data.frame(Year = train_data$Year, GII = train_data$GII, Type = "Actual"),
  data.frame(Year = train_data$Year, GII = train_data$Predicted, Type = "Predicted"),
  data.frame(Year = future_years$Year, GII = future_years$Predicted, Type = "Forecast")
)

# Plot the results
library(ggplot2)
ggplot(combined_data, aes(x = Year, y = GII, color = Type)) +
  geom_line(size = 1) +
  geom_point(size = 2) +
  labs(
    title = "Neural Network Model for Gender Inequality Index",
    x = "Year",
    y = "Gender Inequality Index"
  ) +
  theme_minimal() +
  scale_color_manual(values = c("Actual" = "blue", "Predicted" = "green", "Forecast" = "red"))
```

Neural Network Model for Gender Inequality Index



```
# Evaluation Metrics for Training Data
#install.packages("Metrics")
library(Metrics)
```

```
## Warning: package 'Metrics' was built under R version 4.4.2
```

```
##
## Attaching package: 'Metrics'
```

```
## The following object is masked from 'package:forecast':
##
##     accuracy
```

```
# Assuming the neural network model has been trained and predictions have been made
```

```
# Function to calculate R-squared
```

```
r_squared <- function(actual, predicted) {  
  ss_total <- sum((actual - mean(actual))^2)  
  ss_residual <- sum((actual - predicted)^2)  
  return(1 - (ss_residual / ss_total))  
}
```

```
# Calculate predicted values on training data
```

```
train_data$Predicted <- predict(nn_model, train_data)
```

```
# Calculate MSE, RMSE, MAE, and R-squared for training data
```

```
mse_train <- mse(train_data$GII, train_data$Predicted)  
rmse_train <- rmse(train_data$GII, train_data$Predicted)  
mae_train <- mae(train_data$GII, train_data$Predicted)  
r2_train <- r_squared(train_data$GII, train_data$Predicted)
```

```
cat("Training Data Metrics:\n")
```

```
## Training Data Metrics:
```

```
cat("MSE:", mse_train, "\n")
```

```
## MSE: 0.09393184
```

```
cat("RMSE:", rmse_train, "\n")
```

```
## RMSE: 0.306483
```

```
cat("MAE:", mae_train, "\n")
```

```
## MAE: 0.2546226
```

```
cat("R-squared:", r2_train, "\n")
```

```
## R-squared: -101.4518
```

```
# Predict on test data
```

```
test_data$Predicted <- predict(nn_model, test_data)
```

```
# Calculate MSE, RMSE, MAE, and R-squared for test data
```

```
mse_test <- mse(test_data$GII, test_data$Predicted)  
rmse_test <- rmse(test_data$GII, test_data$Predicted)  
mae_test <- mae(test_data$GII, test_data$Predicted)  
r2_test <- r_squared(test_data$GII, test_data$Predicted)
```

```
cat("\nTest Data Metrics:\n")
```



```
##
```

```
## Test Data Metrics:
```

```
cat("MSE:", mse_test, "\n")
```

```
## MSE: 0.0468334
```

```
cat("RMSE:", rmse_test, "\n")
```

```
## RMSE: 0.2164103
```

```
cat("MAE:", mae_test, "\n")
```

```
## MAE: 0.2009875
```

```
cat("R-squared:", r2_test, "\n")
```

```
## R-squared: -202.7696
```