



Northeastern University
College of Professional Studies

ALY6030 : Analysis of Cryptocurrencies

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Introduction

This file contains the R code used to Clean and create visualizations of the dataset.

Loading Libraries

Data Cleaning

Analysis

Overview of Dataset

Initial Analysis is performed to better understand the dataset and variables under observation, the HEAD and TAIL of the dataset, and the structure of dataset is looked-up.

```
df<-raw_data
# Head,Tail and Structure of Dataset
head(df)
```

```
##      id      name symbol num_market_pairs      date_added  max_supply
## 1      1  Bitcoin   BTC           9431 2013-04-28T00:00:00.000Z    21000000
## 2 1027 Ethereum   ETH           5715 2015-08-07T00:00:00.000Z     -100000
## 3  825   Tether   USDT          33404 2015-02-25T00:00:00.000Z     -100000
## 4 3408 USD Coin   USDC           3978 2018-10-08T00:00:00.000Z     -100000
## 5 1839      BNB    BNB            848 2017-07-25T00:00:00.000Z   165116760
## 6   52      XRP    XRP            721 2013-08-04T00:00:00.000Z 1000000000000
## circulating_supply total_supply cmc_rank      price  volume_24h
## 1          19041375      19041375         1 2.975836e+04 33050126432
## 2          120800743      120800743         2 2.015403e+03 20005339261
## 3          75752120651  79713622671         3 9.990728e-01 62815377734
## 4          51002906718  51002906718         4 9.998124e-01 6456376506
## 5          163276975      163276975         5 2.970114e+02 1734215569
## 6          48343101197  99989535142         6 4.164708e-01 1742870711
## volume_change_24h percent_change_1h percent_change_24h percent_change_7d
## 1          16.9329      -1.11925016      -1.33688771      -9.71992518
## 2          41.6235      -1.46387163      -2.58214902     -16.01911485
## 3          10.8443       0.00717129       0.00531811      -0.09199020
```

```

## 4      27.4036      -0.06923264      -0.04835565      0.03524228
## 5       2.0641      -1.57199529      -1.10532376     -10.40013281
## 6       7.7975      -1.42418103      -2.95732608     -22.27725638
##  percent_change_30d percent_change_60d percent_change_90d  market_cap
## 1      -26.39118788      -27.35857161      -32.66720536  566640002997
## 2      -33.67710395      -28.11097724      -35.11523602  243462140262
## 3       -0.11774563       -0.13237996       -0.14339447   75681882724
## 4       0.04125619       -0.00305729       -0.02917071   50993337983
## 5      -28.45273254      -23.93411936      -30.74394665   48495114806
## 6      -46.43812985      -47.51725263      -49.82558922  20133491375
##  market_cap_dominance fully_diluted_market_cap mineable exchange payments
## 1          44.3870          624925461682      Yes      No      No
## 2          19.0580          243462140262      Yes      No      No
## 3           5.9243          79639711610       No      No      Yes
## 4           3.9917          50993337983       No      Yes     No
## 5           3.7991          49041551944       No      Yes     Yes
## 6           1.5789          41647082782       No      Yes     No

```

```
tail(df)
```

```

##      id      name  symbol num_market_pairs      date_added
## 4995  7669      UNCL   UNCL              3 2020-11-13T00:00:00.000Z
## 4996 14809      Zada   ZADA              1 2021-11-19T03:25:19.000Z
## 4997 17800  Shintama SHINTAMA            4 2022-02-02T02:54:25.000Z
## 4998 10556  B.Protocol BPRO              9 2021-06-21T00:00:00.000Z
## 4999 19906  ChoccySwap  CCY              1 2022-05-03T03:53:12.000Z
## 5000 12698  Ninja Protocol  NINJA            4 2021-10-14T02:52:18.000Z
##      max_supply circulating_supply total_supply cmc_rank      price
## 4995  1.7034e+05              0 0.000000e+00      4995  2.166746e+01
## 4996  1.0000e+12              0 1.000000e+12      4996  1.146176e-06
## 4997 -1.0000e+05              0 0.000000e+00      4997  1.959466e-17
## 4998  1.0000e+07              0 2.194432e+06      4998  1.042482e+00
## 4999 -1.0000e+05              0 0.000000e+00      4999  1.505606e-02
## 5000  5.0000e+07              0 5.000000e+07      5000  2.211259e-02
##      volume_24h volume_change_24h percent_change_1h percent_change_24h
## 4995   19801.88         -5.0042      -0.51192217      -3.930204
## 4996   19772.76        -27.5065      -0.02239866      -2.150332

```

```
## 4997 19756.34 -14.0515 -1.33605640 -2.562610
## 4998 19747.28 -23.1379 -2.37697824 -3.786820
## 4999 19710.65 29.5415 -3.76718119 5.643916
## 5000 19688.05 130.7254 -0.10566668 8.525406
## percent_change_7d percent_change_30d percent_change_60d percent_change_90d
## 4995 -26.36190 -36.48762 -18.52411 -41.98006
## 4996 -27.70020 -55.01843 -70.21234 -78.57549
## 4997 -26.30816 -99.99998 -92.93683 -91.70596
## 4998 -17.43283 -73.75611 -74.00641 -81.44987
## 4999 -26.85643 -25.89154 -25.89154 -25.89154
## 5000 -69.30090 -83.46014 -88.17120 -91.29151
## market_cap market_cap_dominance fully_diluted_market_cap mineable exchange
## 4995 0 0 3690836 No No
## 4996 0 0 1146176 No No
## 4997 0 0 0 No No
## 4998 0 0 10424823 No No
## 4999 0 0 0 No No
## 5000 0 0 1105629 No No
## payments
## 4995 No
## 4996 No
## 4997 No
## 4998 No
## 4999 No
## 5000 No
```

```
str(df)
```

```
## 'data.frame': 5000 obs. of 24 variables:
## $ id : int 1 1027 825 3408 1839 52 2010 5426 4687 74 ...
## $ name : chr "Bitcoin" "Ethereum" "Tether" "USD Coin" ...
## $ symbol : chr "BTC" "ETH" "USDT" "USDC" ...
## $ num_market_pairs : int 9431 5715 33404 3978 848 721 440 310 3699 473 ...
## $ date_added : chr "2013-04-28T00:00:00.000Z" "2015-08-07T00:00:00.000Z" "2015-02-25T00:00:00.000Z" ...
## $ max_supply : num 21000000 -100000 -100000 -100000 165116760 ...
## $ circulating_supply : num 1.90e+07 1.21e+08 7.58e+10 5.10e+10 1.63e+08 ...
## $ total_supply : num 1.90e+07 1.21e+08 7.97e+10 5.10e+10 1.63e+08 ...
```

```

## $ cmc_rank          : int  1 2 3 4 5 6 7 8 9 10 ...
## $ price             : num  2.98e+04 2.02e+03 9.99e-01 1.00 2.97e+02 ...
## $ volume_24h        : num  3.31e+10 2.00e+10 6.28e+10 6.46e+09 1.73e+09 ...
## $ volume_change_24h : num  16.93 41.62 10.84 27.4 2.06 ...
## $ percent_change_1h : num  -1.11925 -1.46387 0.00717 -0.06923 -1.572 ...
## $ percent_change_24h : num  -1.33689 -2.58215 0.00532 -0.04836 -1.10532 ...
## $ percent_change_7d  : num  -9.7199 -16.0191 -0.092 0.0352 -10.4001 ...
## $ percent_change_30d : num  -26.3912 -33.6771 -0.1177 0.0413 -28.4527 ...
## $ percent_change_60d : num  -27.35857 -28.11098 -0.13238 -0.00306 -23.93412 ...
## $ percent_change_90d : num  -32.6672 -35.1152 -0.1434 -0.0292 -30.7439 ...
## $ market_cap        : num  5.67e+11 2.43e+11 7.57e+10 5.10e+10 4.85e+10 ...
## $ market_cap_dominance : num  44.39 19.06 5.92 3.99 3.8 ...
## $ fully_diluted_market_cap : num  6.25e+11 2.43e+11 7.96e+10 5.10e+10 4.90e+10 ...
## $ mineable           : chr  "Yes" "Yes" "No" "No" ...
## $ exchange           : chr  "No" "No" "No" "Yes" ...
## $ payments           : chr  "No" "No" "Yes" "No" ...

```

```
describe(df)
```

```

##          vars    n      mean      sd    median
## id          1 5000  9.398840e+03 5.705670e+03    8770.00
## name*       2 5000  2.491370e+03 1.439780e+03    2490.50
## symbol*     3 5000  2.306580e+03 1.332990e+03    2306.50
## num_market_pairs 4 5000  2.804000e+01 5.179600e+02      5.00
## date_added*  5 5000  1.236120e+03 7.803500e+02    1069.00
## max_supply   6 5000  1.267392e+15 2.680090e+16 100000000.00
## circulating_supply 7 5000  3.121563e+14 1.500169e+16   2436206.35
## total_supply  8 5000  1.457883e+16 9.759416e+17 135572005.60
## cmc_rank     9 5000  2.500500e+03 1.443520e+03    2500.50
## price       10 5000  2.022300e+02 4.932980e+03      0.02
## volume_24h   11 5000  3.845968e+11 1.777213e+13   84622.07
## volume_change_24h 12 5000  3.628867e+08 1.327719e+10      0.00
## percent_change_1h 13 5000 -3.800000e-01 9.930000e+00     -0.57
## percent_change_24h 14 5000  3.180000e+00 1.489100e+02     -1.37
## percent_change_7d 15 5000 -1.513000e+01 3.011200e+02    -22.54
## percent_change_30d 16 5000  6.276551e+04 3.707133e+06    -45.10
## percent_change_60d 17 5000  2.578390e+03 1.837644e+05    -45.75

```

## percent_change_90d	18	5000	1.211928e+04	6.835482e+05	-57.29
## market_cap	19	5000	2.707402e+08	8.873538e+09	38319.88
## market_cap_dominance	20	5000	2.000000e-02	6.900000e-01	0.00
## fully_diluted_market_cap	21	5000	3.582016e+12	1.697423e+14	5691501.89
## mineable*	22	5000	1.080000e+00	2.700000e-01	1.00
## exchange*	23	5000	1.040000e+00	2.000000e-01	1.00
## payments*	24	5000	1.020000e+00	1.500000e-01	1.00
##			trimmed	mad	min max
## id			9167.76	7085.35	1.00 2.014100e+04
## name*			2490.98	1848.80	1.00 4.985000e+03
## symbol*			2305.29	1711.66	1.00 4.628000e+03
## num_market_pairs			6.50	4.45	1.00 3.340400e+04
## date_added*			1176.34	765.02	1.00 3.016000e+03
## max_supply			838241502.15	148408260.00	-100000.00 1.000000e+18
## circulating_supply			99452414.77	3611919.54	0.00 9.818468e+17
## total_supply			1161578481.88	200999055.50	0.00 6.900000e+19
## cmc_rank			2500.50	1853.25	1.00 5.000000e+03
## price			0.20	0.03	0.00 3.068975e+05
## volume_24h			369164.62	125460.68	0.00 1.068732e+15
## volume_change_24h			5.60	28.60	-100.00 6.136198e+11
## percent_change_1h			-0.63	0.91	-85.50 5.639200e+02
## percent_change_24h			-0.94	3.83	-99.38 1.010783e+04
## percent_change_7d			-22.73	19.01	-100.00 2.058222e+04
## percent_change_30d			-43.11	23.70	-100.00 2.555570e+08
## percent_change_60d			-43.13	27.24	-100.00 1.299406e+07
## percent_change_90d			-53.48	27.99	-100.00 4.626783e+07
## market_cap			2151091.82	56813.05	0.00 5.666400e+11
## market_cap_dominance			0.00	0.00	0.00 4.439000e+01
## fully_diluted_market_cap			30232309.74	8409502.61	0.00 8.824270e+15
## mineable*			1.00	0.00	1.00 2.000000e+00
## exchange*			1.00	0.00	1.00 2.000000e+00
## payments*			1.00	0.00	1.00 2.000000e+00
##			range	skew kurtosis	se
## id			2.014000e+04	0.28 -1.08	8.069000e+01
## name*			4.984000e+03	0.00 -1.20	2.036000e+01
## symbol*			4.627000e+03	0.00 -1.20	1.885000e+01

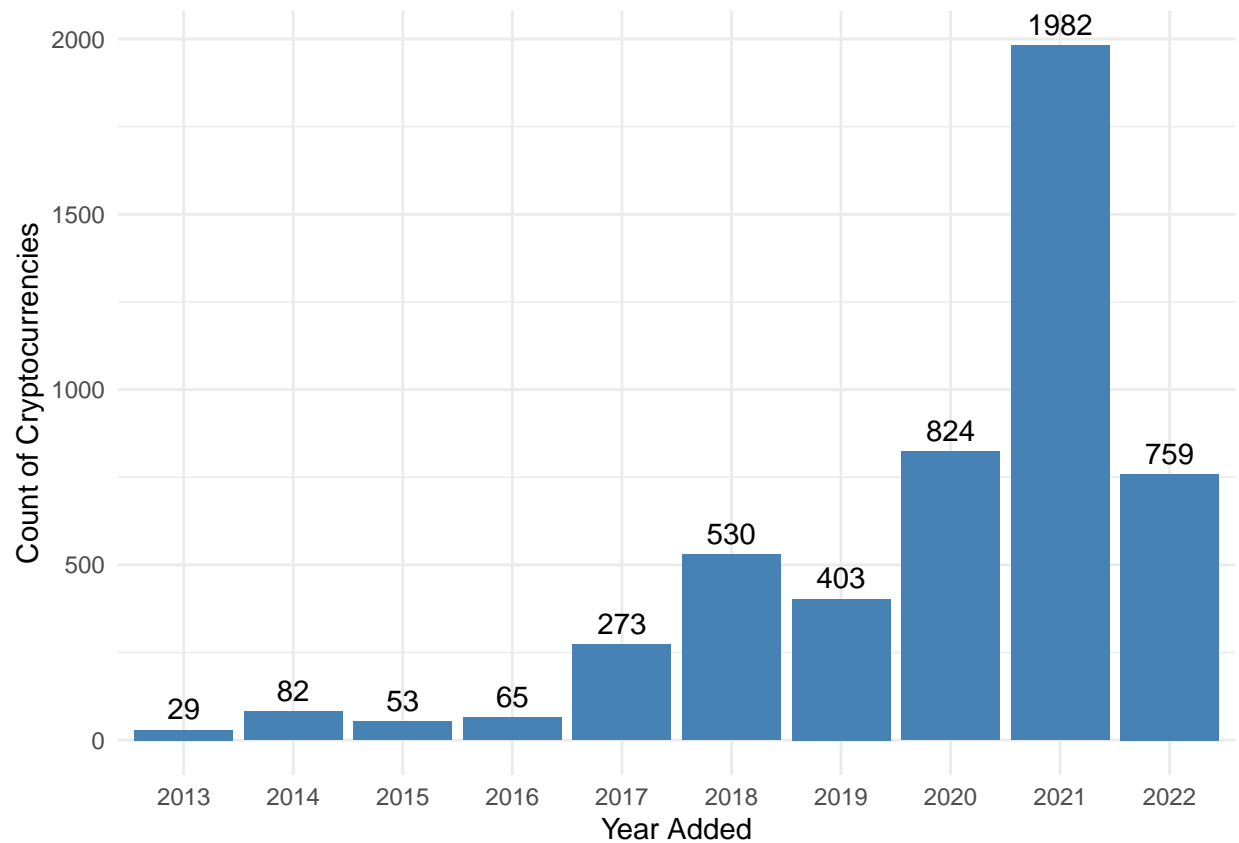
```
## num_market_pairs      3.340300e+04 55.70  3477.44 7.330000e+00
## date_added*           3.015000e+03  0.64    -0.56 1.104000e+01
## max_supply            1.000000e+18 32.86  1179.22 3.790220e+14
## circulating_supply     9.818468e+17 59.63  3756.52 2.121560e+14
## total_supply          6.900000e+19 70.64  4990.03 1.380190e+16
## cmc_rank              4.999000e+03  0.00    -1.20 2.041000e+01
## price                 3.068975e+05 50.94  3033.24 6.976000e+01
## volume_24h            1.068732e+15 53.28  2966.85 2.513358e+11
## volume_change_24h     6.136198e+11 40.63  1698.72 1.877678e+08
## percent_change_1h     6.494200e+02 39.31  2128.83 1.400000e-01
## percent_change_24h    1.020720e+04 63.28  4248.21 2.110000e+00
## percent_change_7d     2.068222e+04 64.50  4380.26 4.260000e+00
## percent_change_30d    2.555571e+08 66.25  4521.61 5.242678e+04
## percent_change_60d    1.299416e+07 70.67  4992.97 2.598820e+03
## percent_change_90d    4.626793e+07 63.68  4225.77 9.666830e+03
## market_cap            5.666400e+11 56.33  3430.55 1.254908e+08
## market_cap_dominance  4.439000e+01 56.48  3443.56 1.000000e-02
## fully_diluted_market_cap 8.824270e+15 49.63  2479.47 2.400518e+12
## mineable*             1.000000e+00  3.16    7.97 0.000000e+00
## exchange*             1.000000e+00  4.54   18.62 0.000000e+00
## payments*             1.000000e+00  6.22   36.68 0.000000e+00
```

```
top_ranked <- head(df)
bottom_ranked <- tail(df)
```

Barplot of Year in which Cryptocurrency was added

```
df$year <- substr(df$date_added, 0, 4)

ggplot(data=df, aes(x=year)) +
  geom_bar(stat="count", fill="steelblue")+
  geom_text(stat='count', aes(label=..count..), vjust=-0.5)+
  xlab("Year Added") + ylab("Count of Cryptocurrencies")+
  theme_minimal()
```

Top 10 Most Expensive Crypto

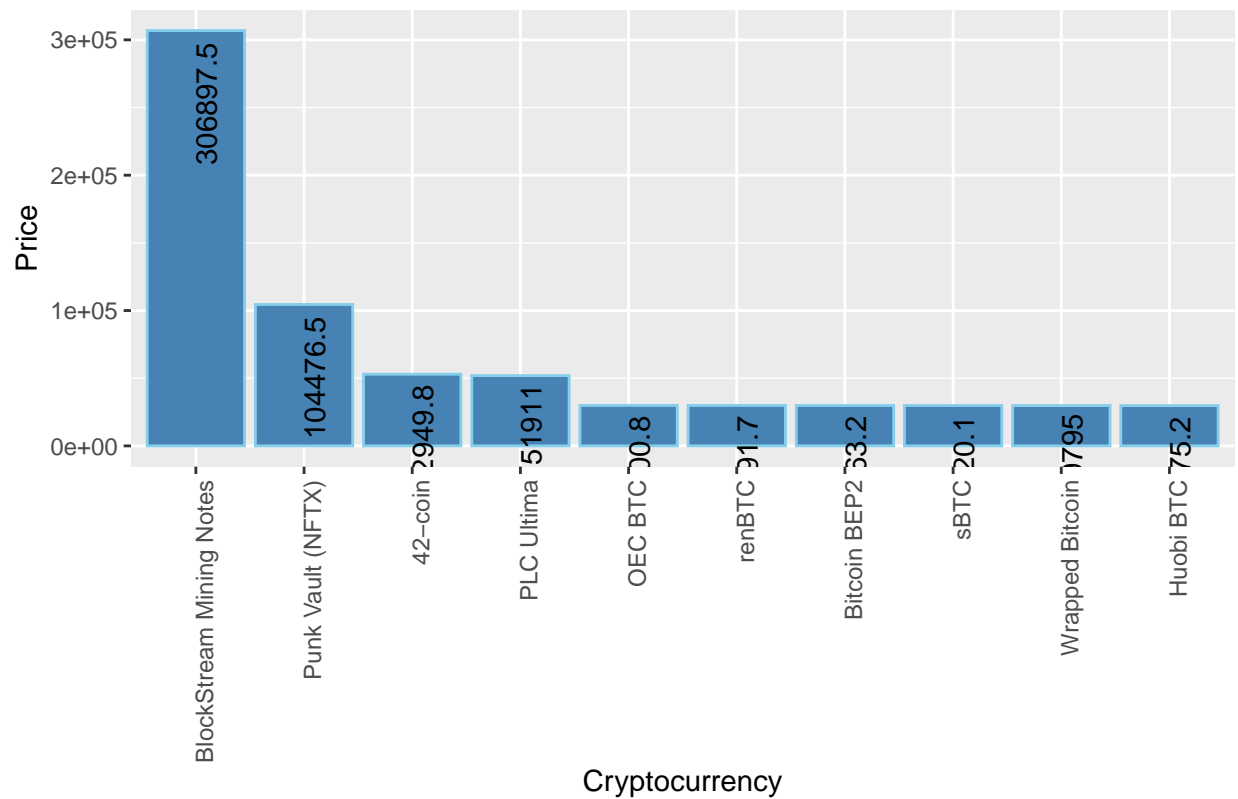
```
## OrderBy Price (DESC)
df2 <- df[order(-df$price),]

## Filter Top 10
priceDesc <- top_n(df2,10,df2$price)

p <- ggplot(priceDesc, aes(x = reorder(name, -price), y = price)) + geom_bar(stat="identity", color='skyblue') +
  labs(y="Price", x="Cryptocurrency", title="Top 10 Most Expensive Crypto")+
  theme(axis.text.x=element_text(angle=90, hjust=1))+
  geom_text(label=round(priceDesc$price,digit=1),position=position_dodge(width=0.3),
            hjust=1.1,angle=90,vjust=1)

p
```

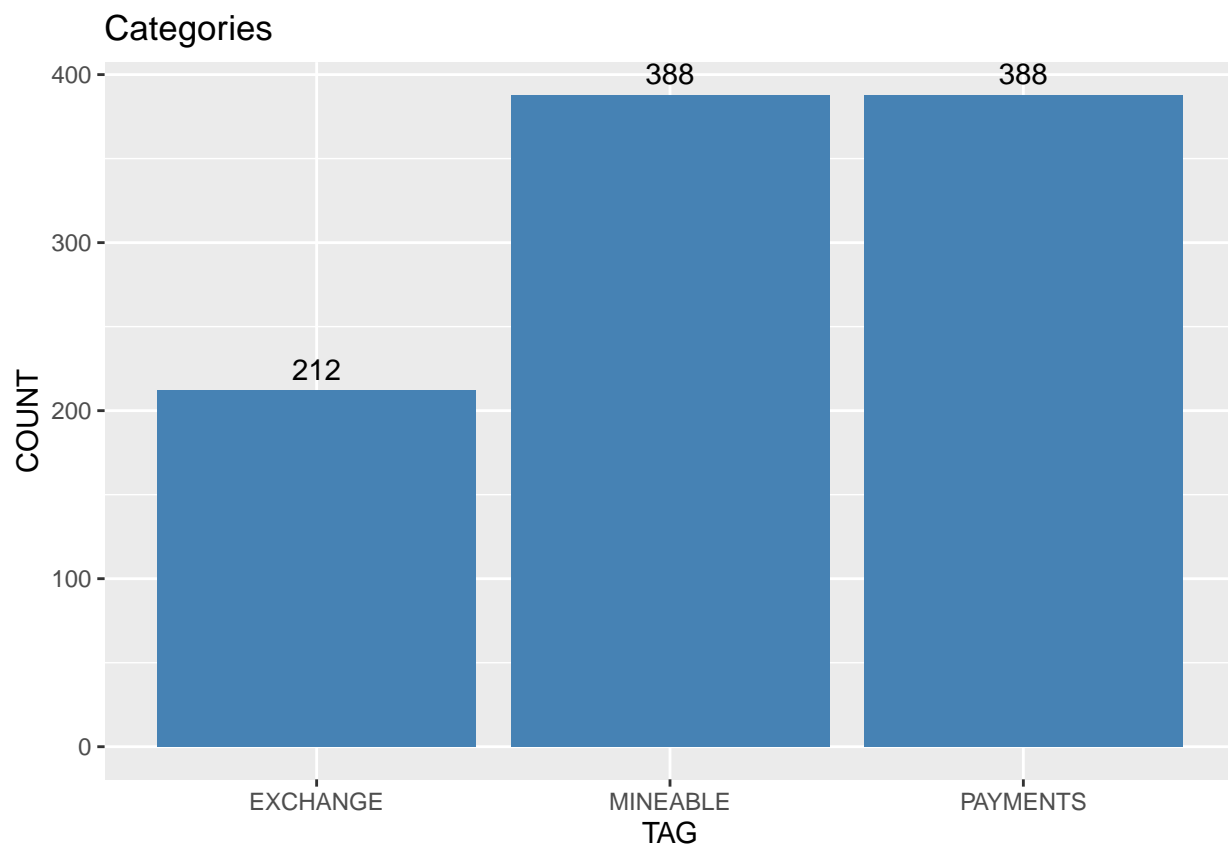
Top 10 Most Expensive Crypto



Count of Tags

```
catCount <-c()
for (i in 1:nrow(df))
{
  if(identical(df[i,22], "Yes")){
    catCount <- append(catCount, "MINEABLE")
  }
  if(identical(df[i,23], "Yes")){
    catCount <- append(catCount, "EXCHANGE")
  }
  if(identical(df[i,22], "Yes")){
    catCount <- append(catCount, "PAYMENTS")
  }
}
```

```
q <- ggplot(data.frame(catCount), aes(x=catCount)) +
  geom_bar(stat="count", fill="steelblue")+
  geom_text(stat='count', aes(label=..count..), vjust=-0.5)+
  labs(y="COUNT", x="TAG", title="Categories")
q
```

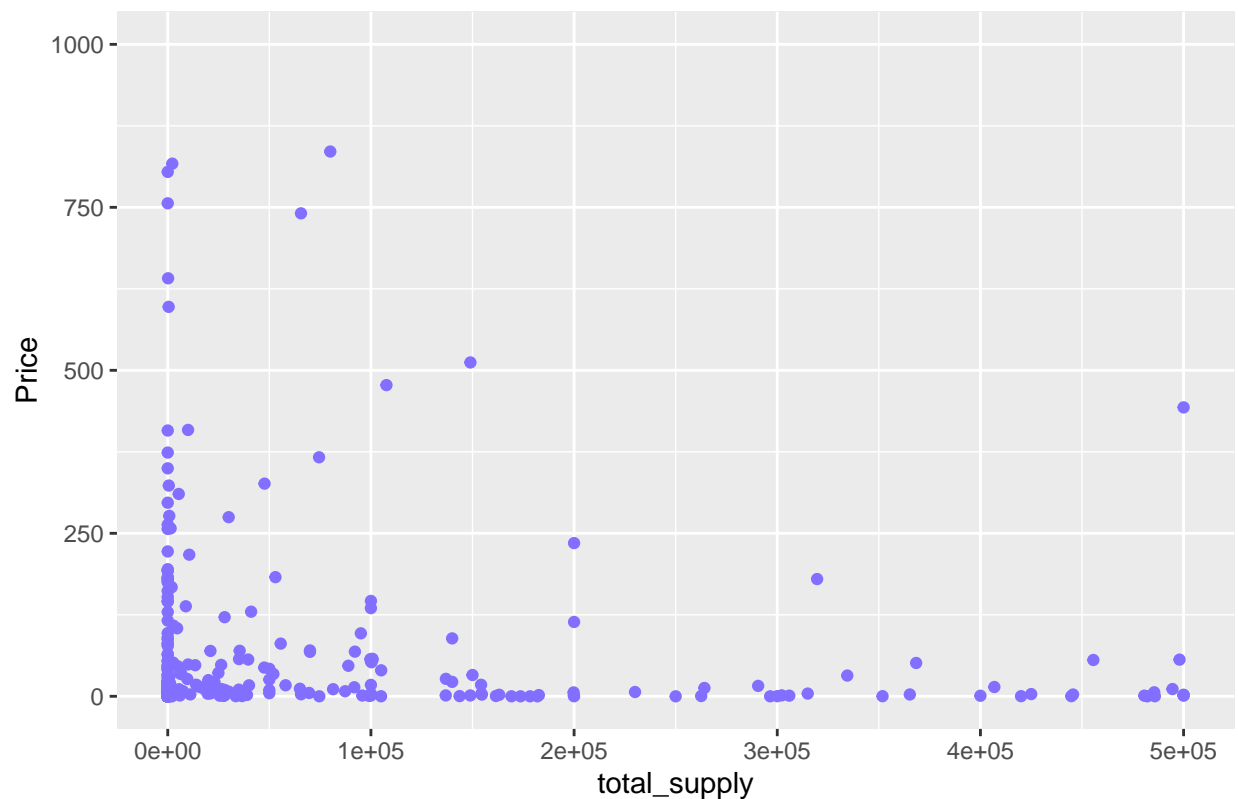


Scatterplot of Price Vs Total Supply of currencies in Lower Price Range

```
ggplot(data=df, aes(total_supply, price)) +
  geom_point(color="slateblue1") +
  ggtitle("Scatterplot of Price Vs Total Supply of currencies in Lower Price Range") + labs(y="Price", x="Total Supply")
```

Warning: Removed 4164 rows containing missing values (geom_point).

Scatterplot of Price Vs Total Supply of currencies in Lower Price Range

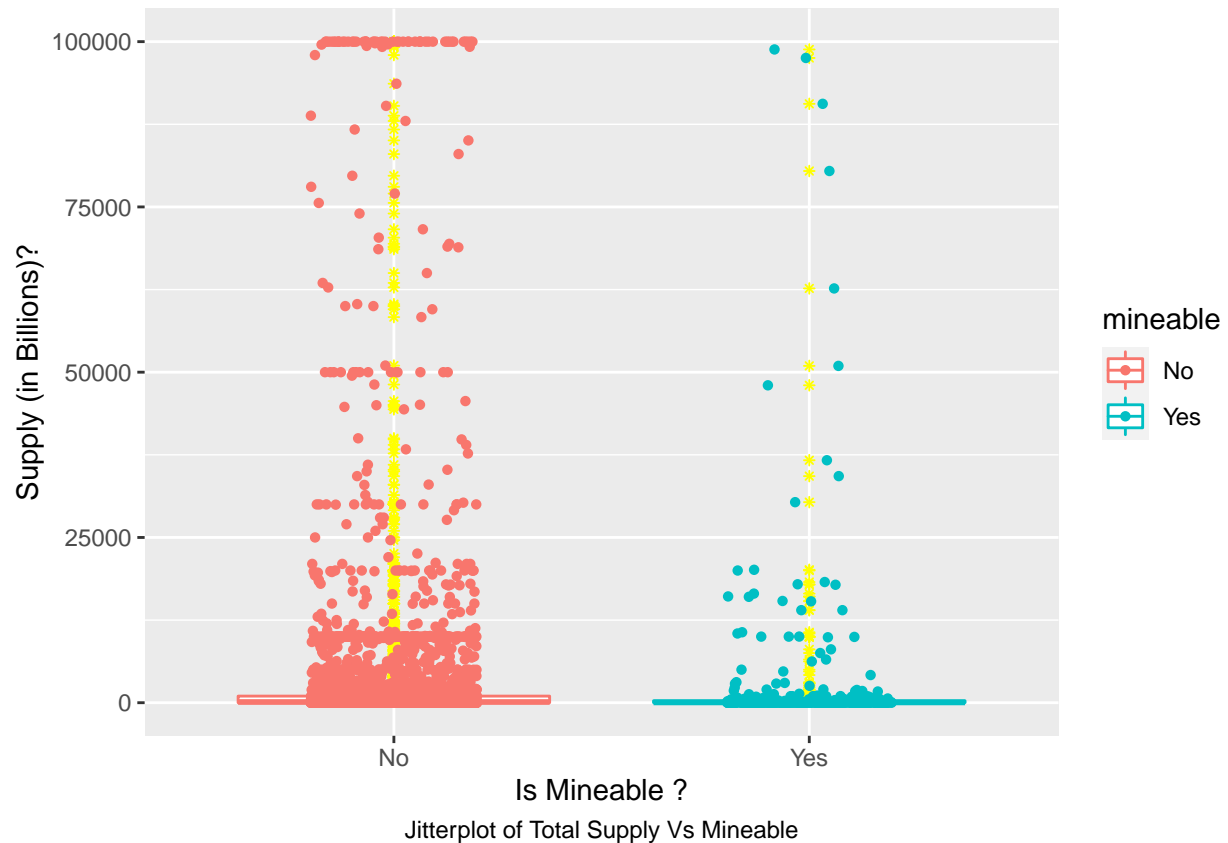


Jitterplot of Total Supply Vs Mineable

```
ggplot(df, aes(x=mineable, y=total_supply/1000000, color=mineable)) +
  geom_boxplot(notch=FALSE, outlier.colour="yellow", outlier.shape=8,
              outlier.size=1) +
  geom_jitter(shape=16, position=position_jitter(0.2)) +
  labs(caption="Jitterplot of Total Supply Vs Mineable",
       x="Is Mineable ?", y = "Supply (in Billions)?") +
  theme(plot.caption = element_text(hjust = 0.5))+ylim(c(0,100000))
```

Warning: Removed 256 rows containing non-finite values (stat_boxplot).

Warning: Removed 613 rows containing missing values (geom_point).



ANOVA Test

```
# One-way ANOVA Test

# Set significance level
alpha <- 0.05

# Dataframe for 30 Days
thirtyDays <- data.frame('variation' = df$percent_change_30d,
                          'timeperiod' = rep('thirtyDays',5000), stringsAsFactors = FALSE)

# Dataframe for 60 Days
sixtyDays <- data.frame('variation' = df$percent_change_60d,
                        'timeperiod' = rep('sixtyDays',5000), stringsAsFactors = FALSE)

# Dataframe for 90 Days
ninetyDays <- data.frame('variation' = df$percent_change_90d,
                         'timeperiod' = rep('ninetyDays',5000), stringsAsFactors = FALSE)

# Combine the Dataframe
```

```

variation <- rbind(thirtyDays,sixtyDays,ninetyDays)
variation$timeperiod <- as.factor(variation$timeperiod)

# Hypotheses
# H0: Mean Variation(30 Days) = Mean Variation(60 Days) = Mean Variation(90 Days)
# H1: Atleast one mean is different from others

anova <- aov(variation ~ timeperiod, data = variation)
a_summ <-summary(anova)
# Critical Value
qf(1-alpha,a_summ[[1]][1,1],a_summ[[1]][2,1])

```

```
## [1] 2.996331
```

```

# Test Value
F.value <- a_summ[[1]][[1,"F value"]]
F.value

```

```
## [1] 1.101977
```

```

# Compare p-value and alpha to make decision
p.value <- a_summ[[1]][[1,"Pr(>F)"]]
p.value

```

```
## [1] 0.3322404
```

```
ifelse(p.value > alpha,"Failed to reject Null Hypothesis","Reject Null Hypothesis")
```

```
## [1] "Failed to reject Null Hypothesis"
```

Correlation Analysis of Numeric values

```

numeric_data <- select_if(df, is.numeric)
data.cor = cor(numeric_data, method = c("spearman"))
corrplot(data.cor)

```

