

# BT2201 Finance Project

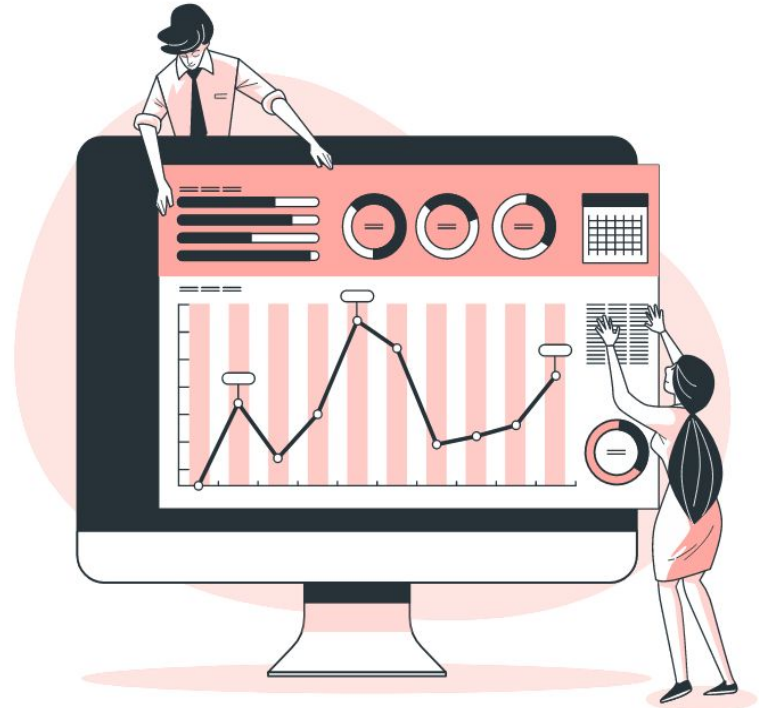
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# 01

## ETF Details



# Considerations: Expense Ratio, Coverage, Holdings

ETF	Benchmark	Expense Ratio/%	Sector	Region	Major Holdings
<b>AGG US</b>	Bloomberg US Aggregate	0.03	Bond	US	BlackRock Funds, US Treasury Notes
<b>SCHP</b>	Bloomberg US Treasury Inflation Protected Notes (TIPS)	0.05	Bond	US	US Treasury Notes
<b>VCSH</b>	Bloomberg US Corporate	0.04	Corporate Bonds	Global Developed Markets	US Dollar, Boeing Company, Apple Inc., Bank of America Corporation
<b>BNDX</b>	Bloomberg Global Aggregate x USD Float Adjusted RIC Capped	0.07	Bond	Global	Japan, France, Germany, Italy government bonds
<b>VWOB</b>	Bloomberg USD Emerging Markets Govt. RIC Capped Bond	0.20	Bond	Global Emerging markets	Qatar, Saudi Arabia, Indonesia
<b>1306 JT</b>	Topix Gross Total Return Index	0.04	Mid/Small Cap	Japan	Toyota Motor Corp, Sony Corp, Softbank Group Corp

# Considerations: Expense Ratio, Coverage, Holdings

ETF	Benchmark	Expense Ratio/%	Sector	Region	Major Holdings
<b>VOO</b> US	S&P 500	0.03%	Large Cap	US	APPL, MSFT, AMZN, TSLA
<b>VO</b> US	CRSP US Mid Cap	0.04	Mid- Cap	US	Palo Alto Networks, Pioneer Natural Resources Company
<b>VSS</b>	FTSE Global ex Small Cap Net Tax(US RIC) Index	0.07	Large/Mid-Cap/ Small	Global	US Dollar, Vanguard Cash Management, First Quantum Minerals
<b>VGK</b>	FTSE Developed Europe All Cap Net Tax(US RIC) Index	0.08	Large-Cap	Europe	Nestle SA, Roche Holding AG, Novartis AG
<b>VWO</b>	FTSE Custom Emerging Markets All Cap China A Inclusion Net Tax (US RIC)	0.08	Large-Cap	Global Emerging Markets	Taiwan SemiConductor, Tencent Holdings, Alibaba
<b>1343</b> JT	TSE REIT Index	0.12	REITS	Japan	Nippon Building Fund, Japan REIT Group, Japan Metropolitan Fund Inv.

# Considerations: Expense Ratio, Coverage, Holdings

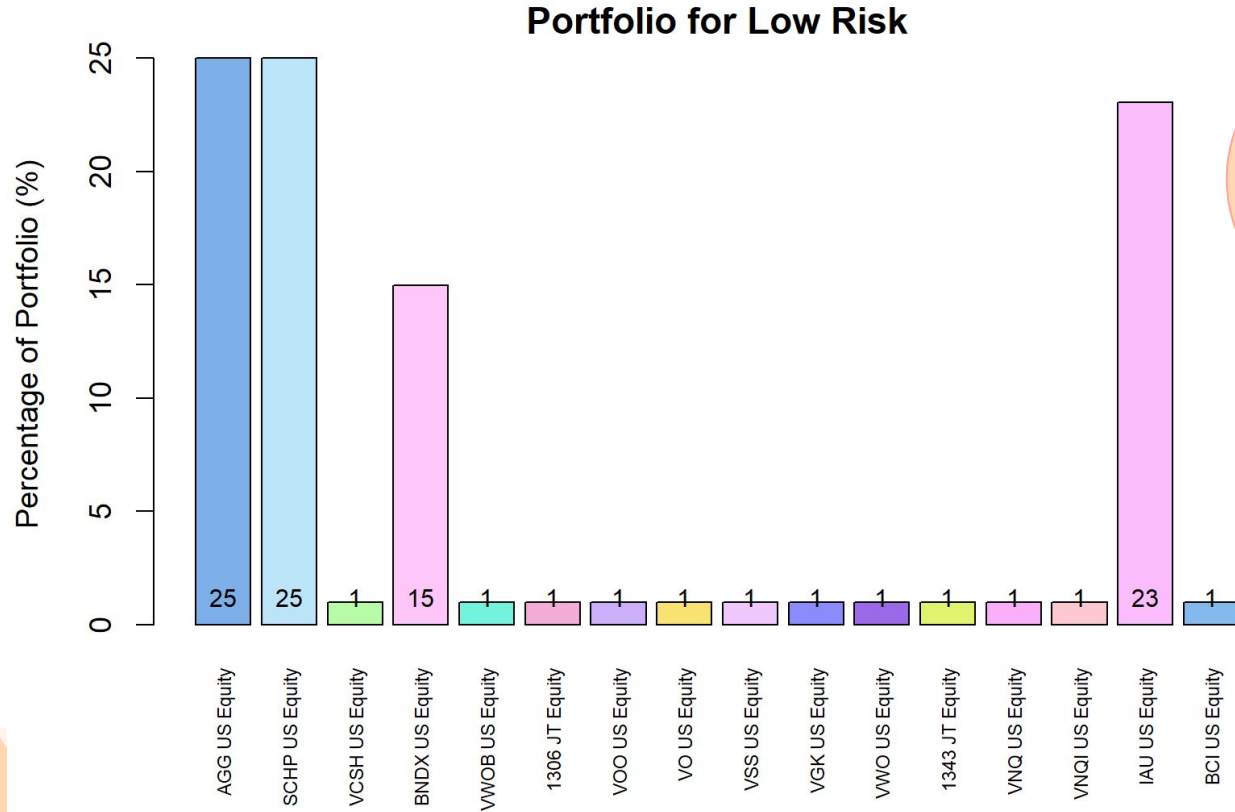
ETF	Benchmark	Expense Ratio/%	Sector	Region	Major Holdings
<b>VNQ US</b>	MSCI US Investable Market Real Estate 25/50	0.12	REIT	US	Vanguard Real Estate II Index Fund, Prologis Inc, American Tower Corporation
<b>VNQI US</b>	S&P Global ex-US Property Index	0.12	Large/Mid-Cap REITS	Global	Vonovia SE, Goodman Group, Mitsui Fudosan Co. , Mitsubishi Estate Company
<b>IAU US</b>	LBMA Gold Price PM	0.25	Commodity	-	Gold
<b>BCI</b>	Aberdeen Bloomberg All Commodity ETF	0.25	Commodity	Global	Gold FUTR, Crude Future, Natural Gas Future, Corn Future

# 02



## Overview of Portfolios

# Low Risk



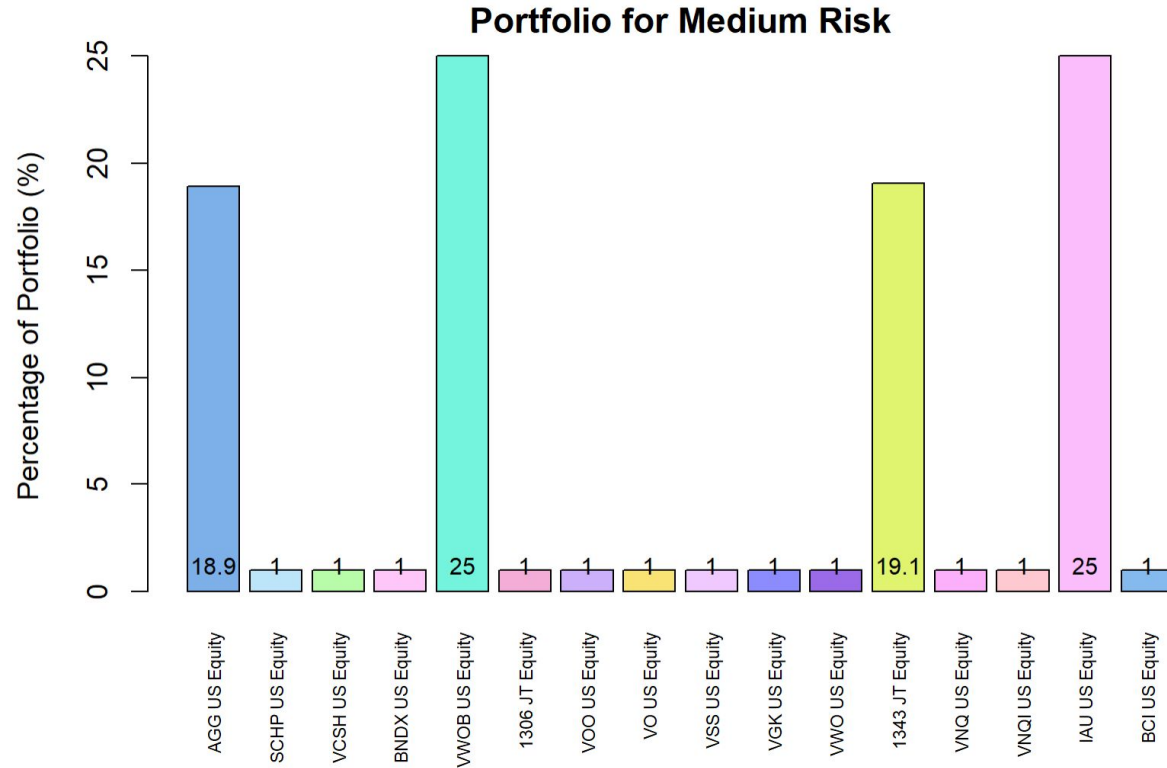
**Bonds: 67%**  
**Equities: 9%**  
**Gold &  
Commodities: 24%**



# Low Risk Portfolio

Annual expected returns	Volatility	Sharpe Ratio
5.75%	0.105	0.535

# Medium Risk

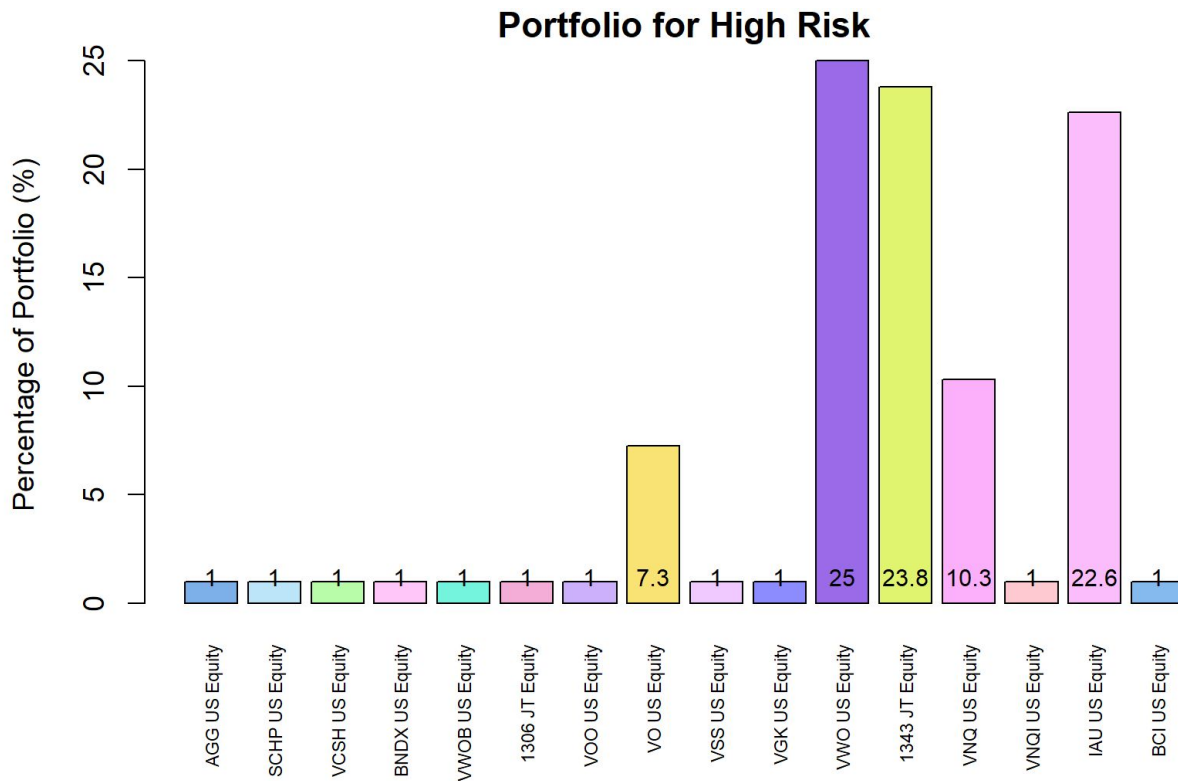


**Bonds: 46.9%**  
**Equities: 27.1%**  
**Gold & Commodities: 26%**

# Medium Risk Portfolio

Annual expected returns	Volatility	Sharpe Ratio
8.08%	0.129	0.613

# High Risk



**Bonds: 5%**  
**Equities: 71.4%**  
**Gold & Commodities: 23.6%**

# High Risk Portfolio

Annual expected returns	Volatility	Sharpe Ratio
11.0%	0.193	0.562



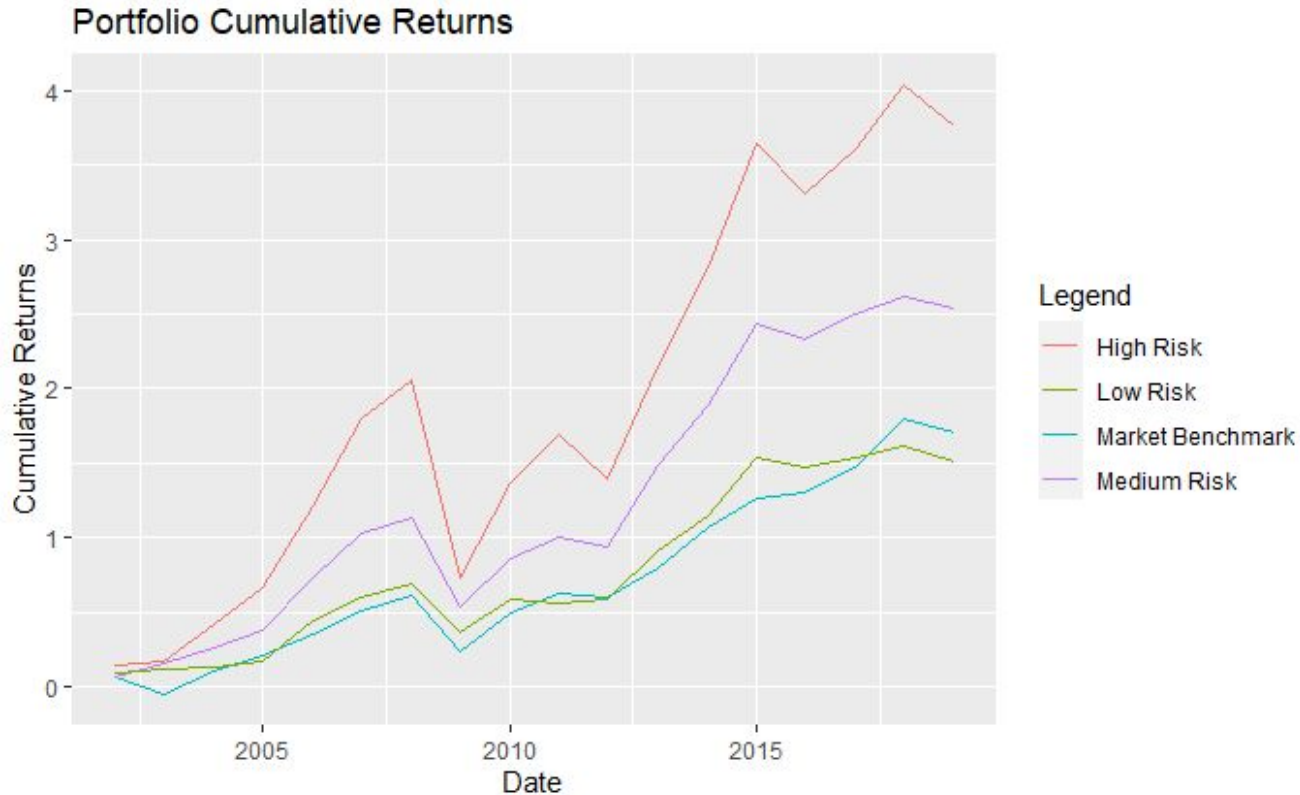
# 03

**Performance  
against  
market**

# Market Performance

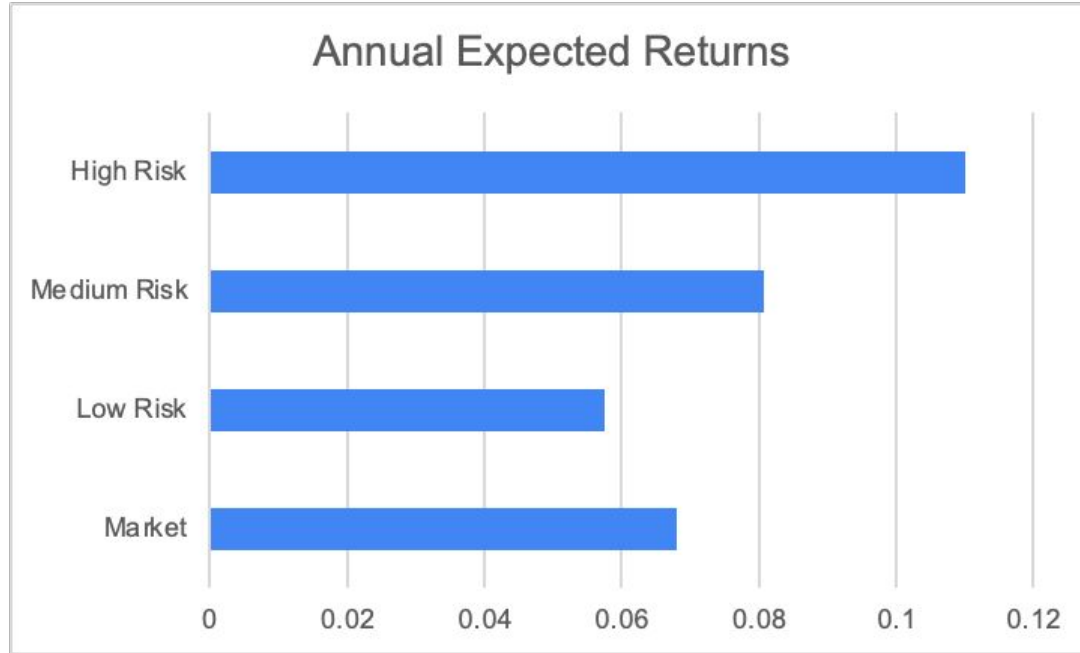
Annual expected returns	Volatility	Sharpe Ratio
6.23%	0.106	0.576

# Benchmark against market portfolio

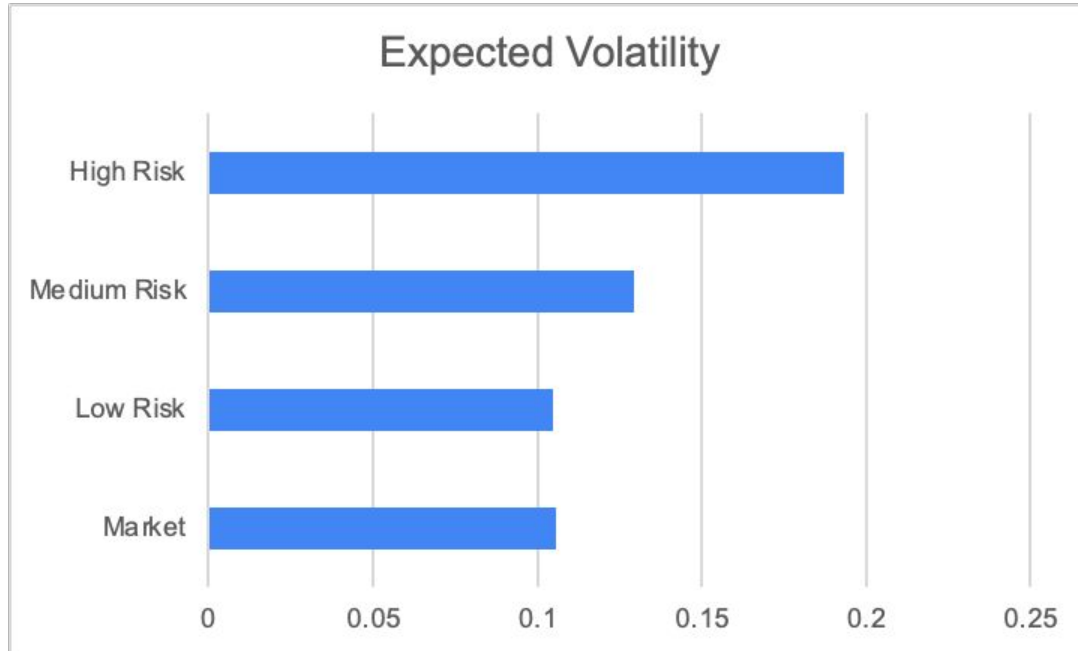




# Comparing returns



# Comparing volatility



## Low Risk Portfolio Beta

Call:

```
lm(formula = minvarreturns ~ testfull$mkt)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.119500	-0.049629	-0.000753	0.026188	0.143857

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.01368	0.02096	0.653	0.523298
testfull\$mkt	<u>0.70426</u>	0.17471	4.031	<u>0.000967</u> ***

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Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.07601 on 16 degrees of freedom

Multiple R-squared: 0.5039, Adjusted R-squared: 0.4728

F-statistic: 16.25 on 1 and 16 DF, p-value: 0.000967

## High Risk Portfolio Beta

```
Call:
lm(formula = maxreturnreturns ~ testfull$mkt)

Residuals:
    Min       1Q   Median       3Q      Max
-0.13054 -0.06385 -0.01047  0.05356  0.19994

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.008447   0.024979   0.338    0.74
testfull$mkt  1.631314   0.208228   7.834 7.27e-07 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.09059 on 16 degrees of freedom
Multiple R-squared:  0.7932,    Adjusted R-squared:  0.7803
F-statistic: 61.38 on 1 and 16 DF,  p-value: 7.268e-07
```

## Medium Risk Portfolio Beta

Call:

```
lm(formula = tangencyreturns ~ testfull$mkt)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.11401	-0.04355	-0.01559	0.03040	0.16617

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.02053	0.02261	0.908	0.377
testfull\$mkt	<u>0.96806</u>	0.18847	5.136	9.95e-05 ***

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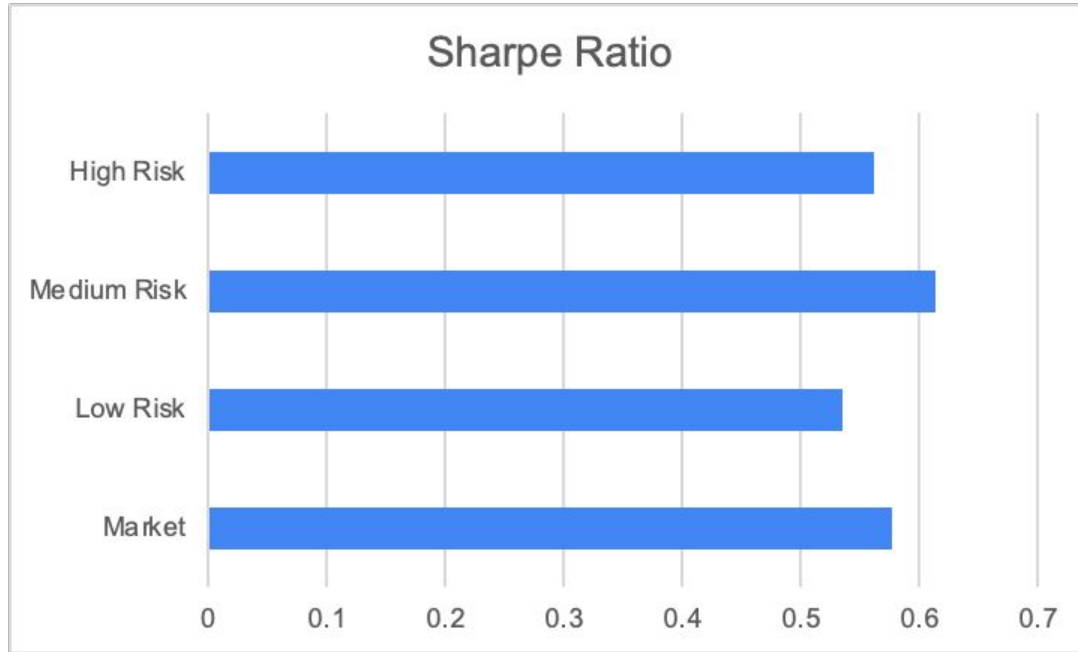
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Residual standard error: 0.082 on 16 degrees of freedom

Multiple R-squared: 0.6225, Adjusted R-squared: 0.5989

F-statistic: 26.38 on 1 and 16 DF, p-value: 9.95e-05

# Comparing Sharpe Ratios





# 04

## Appendix

# Portfolio Construction Process

- Data cleaning and manipulation
  - To account for the exchange rate JPYUSD changes, taking into consideration that there are US ETFs in the portfolio, we mutated the daily returns of each US ETF through the equation:

$$\text{Currency Accounted returns} = (1 + \text{daily returns}) * (1 + \text{JPYUSD returns}) - 1$$

- To manipulate our data for portfolio optimisation in R, we then converted our returns into an time series object, which then allowed us to find the yearly geometric returns for each ETF in the portfolio.



# Portfolio Construction Process

	mkt		rf		AGG	US Equity	SCHP	US Equity
2001-12-31	0.06649724	0.0001463167	0.101706546	0.08935238				
2002-12-31	-0.11215371	0.0005551092	-0.005226801	0.05171745				
2003-12-31	0.16839124	0.0004120871	-0.061233960	-0.02156482				
2004-12-31	0.09251594	0.0003751140	-0.006588657	0.03815481				
2005-12-30	0.12074471	0.0004121510	0.173296013	0.17990141				
2006-12-29	0.11611596	0.0021551581	0.050515771	0.01529592				
	VCSH		US Equity		BNDX	US Equity	VWOB	US Equity
2001-12-31	0.19959965	0.10717871	0.14101733	0.01176471				
2002-12-31	-0.05108981	-0.03598708	0.02820919	-0.17829457				
2003-12-31	-0.11686321	-0.07552222	0.13680720	0.25613208				
2004-12-31	-0.00628774	0.00752676	0.07004455	0.10918137				
2005-12-30	0.35373079	0.20949474	0.27083002	0.45315277				
2006-12-29	0.02659136	0.04341680	0.11486169	0.03127730				
	VVO		US Equity		VO	US Equity	VSS	US Equity
2001-12-31	0.2188594	0.2992020	0.1800537	0.2047633				
2002-12-31	-0.2971264	-0.2286476	-0.1603647	-0.2736456				
2003-12-31	0.1613335	0.2239772	0.4681771	0.2214153				
2004-12-31	0.0612918	0.1493192	0.2445964	0.1279667				
2005-12-30	0.2036238	0.3075809	0.4433798	0.2532602				
2006-12-29	0.1706475	0.1519765	0.2124201	0.3454118				
	VWO		US Equity		1343	JT Equity	VNQ	US Equity
2001-12-31	0.3974656	-0.0748600	0.15557763	0.14672958				
2002-12-31	-0.1513569	0.2739261	-0.06486847	-0.08918043				
2003-12-31	0.4064956	0.1637695	0.23425166	0.31111881				
2004-12-31	0.2061741	0.3197815	0.25572406	0.33428380				
2005-12-30	0.5231878	0.1208753	0.28431062	0.34327039				
2006-12-29	0.3080363	0.2899283	0.36802756	0.46793148				
	VNQI		US Equity		VNQI	US Equity		

# Portfolio Construction Process

- Portfolio construction and optimisation
  - We ran individual regression to obtain individual  $\beta_{OLS}$  of each ETF and find the respective  $\mu_{CAPM}$  and total risk.
  - By having our  $\mu_{CAPM}$  and total risk we are then able to obtain our weights for the portfolio.

$$w^* = f(\mu, \Sigma)$$

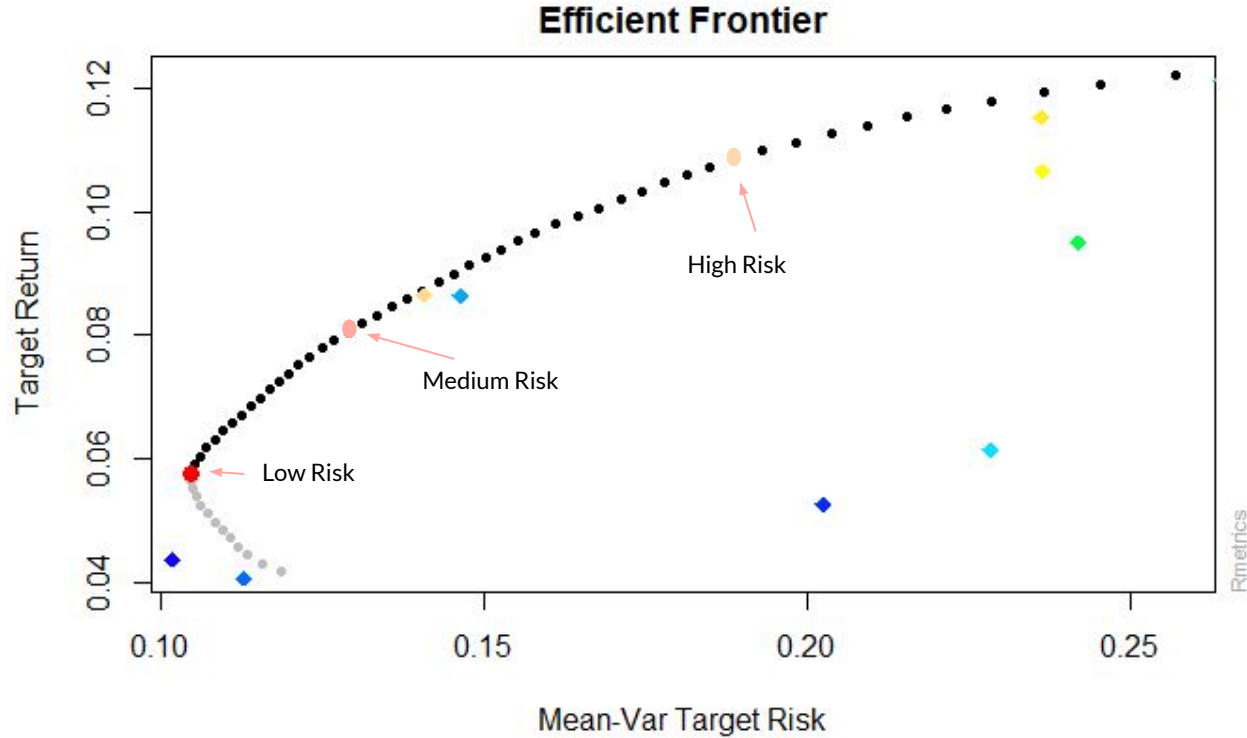
- From the equation, we can use the covariance matrix and the  $\mu_{CAPM}$  to find the weights of our portfolio. However, inspecting the covariance matrix led us to find that some of the ETF returns are highly correlated to each other, making determinant of the matrix to be equal to 0, thus making it non invertible. This prevents us from finding reasonable weights using  $w^* = f(\mu, \Sigma)$ . Thus, we added constraints to include all the ETFs provided.
- This was done through the use of the R packages PerformanceAnalytics and fPortfolio.

# Portfolio Construction Process

- Portfolio construction and optimisation
  - To form our 3 portfolios, we sought for points along the efficient frontier.
  - These are the points which gave the minimum variance portfolio, the tangency portfolio and the maximum returns portfolio.
  - In order to maximise diversification, we have set minimum weights for each ETF in the portfolio to be 1%.
  - Diversification is done to minimise idiosyncratic risk.
  - Ideally, the portfolio constructed should only be left with its systematic risk.
  - Hence, the investor will be compensated with higher return for every unit of risk taken on.

# Portfolio Construction Process

## Efficient Frontier



# Portfolio Construction Process

## Min Variance Portfolio

### Title:

MV Minimum Variance Portfolio  
Estimator: covEstimator  
Solver: solveRquadprog  
Optimize: minRisk  
Constraints: minW maxW

### Portfolio Weights:

AGG US Equity	SCHP US Equity	VCSH US Equity	BNDX US Equity
0.2500	0.1674	0.0100	0.2126
VWOB US Equity	1306 JT Equity	VVO US Equity	VO US Equity
0.0100	0.0100	0.0100	0.0100
VSS US Equity	VGK US Equity	VWO US Equity	1343 JT Equity
0.0100	0.0100	0.0100	0.0100
VNQ US Equity	VNQI US Equity	IAU US Equity	BCI US Equity
0.0100	0.0100	0.2500	0.0100

### Covariance Risk Budgets:

AGG US Equity	SCHP US Equity	VCSH US Equity	BNDX US Equity
0.2020	0.1553	0.0179	0.1973
VWOB US Equity	1306 JT Equity	VVO US Equity	VO US Equity
0.0129	0.0146	0.0189	0.0209
VSS US Equity	VGK US Equity	VWO US Equity	1343 JT Equity
0.0211	0.0201	0.0237	0.0132
VNQ US Equity	VNQI US Equity	IAU US Equity	BCI US Equity
0.0195	0.0225	0.2257	0.0143

### Target Returns and Risks:

mean	Cov	CVaR	VaR
0.0620	0.0964	0.1777	0.0365

# Portfolio Construction Process

## Tangency Portfolio

```
Title:
MV Tangency Portfolio
Estimator:      covEstimator
Solver:         solveRquadprog
Optimize:       minRisk
Constraints:    minW maxW

Portfolio Weights:
AGG US Equity SCHP US Equity VCSH US Equity BNDX US Equity
0.0671      0.2500      0.0100      0.0100
VWOB US Equity 1306 JT Equity VOO US Equity  VO US Equity
0.1622      0.0100      0.0100      0.0572
VSS US Equity  VGK US Equity  VWO US Equity 1343 JT Equity
0.0100      0.0100      0.0100      0.1135
VNQ US Equity  VNQI US Equity IAU US Equity  BCI US Equity
0.0100      0.0100      0.2500      0.0100

Covariance Risk Budgets:
AGG US Equity SCHP US Equity VCSH US Equity BNDX US Equity
0.0395      0.1803      0.0133      0.0068
VWOB US Equity 1306 JT Equity VOO US Equity  VO US Equity
0.1779      0.0125      0.0159      0.1029
VSS US Equity  VGK US Equity  VWO US Equity 1343 JT Equity
0.0195      0.0176      0.0211      0.1542
VNQ US Equity  VNQI US Equity IAU US Equity  BCI US Equity
0.0174      0.0206      0.1880      0.0127

Target Returns and Risks:
mean    Cov    CVaR    VaR
0.0796  0.1173  0.2544  0.0382
```

# Portfolio Construction Process

## Max Returns Portfolio

### Title:

MV Return Maximized Efficient Portfolio

Estimator: covEstimator

Solver: solveRquadprog

Optimize: minRisk

Constraints: minW maxW

### Portfolio Weights:

AGG US Equity	SCHP US Equity	VCSH US Equity	BNDX US Equity
0.0100	0.0100	0.0100	0.0100
VWOB US Equity	1306 JT Equity	VVO US Equity	VO US Equity
0.0410	0.0100	0.0100	0.2500
VSS US Equity	VGK US Equity	VWO US Equity	1343 JT Equity
0.0100	0.0100	0.0365	0.1195
VNQ US Equity	VNQI US Equity	IAU US Equity	BCI US Equity
0.2029	0.0100	0.2500	0.0100

### Covariance Risk Budgets:

AGG US Equity	SCHP US Equity	VCSH US Equity	BNDX US Equity
0.0038	0.0045	0.0087	0.0044
VWOB US Equity	1306 JT Equity	VVO US Equity	VO US Equity
0.0298	0.0094	0.0121	0.3385
VSS US Equity	VGK US Equity	VWO US Equity	1343 JT Equity
0.0143	0.0130	0.0529	0.1166
VNQ US Equity	VNQI US Equity	IAU US Equity	BCI US Equity
0.2648	0.0145	0.1042	0.0084

### Target Returns and Risks:

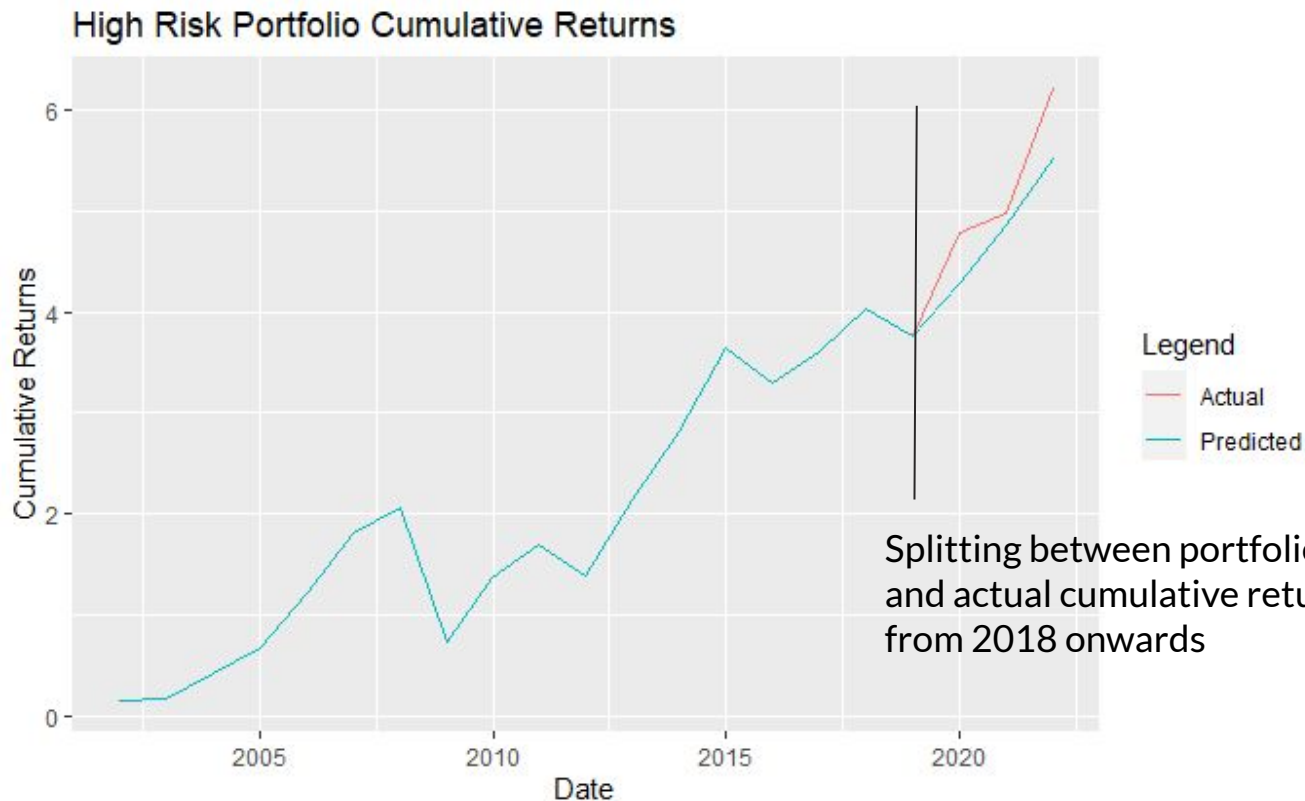
mean	Cov	CVaR	VaR
0.1100	0.1710	0.3690	0.0631

# Portfolio Construction Process

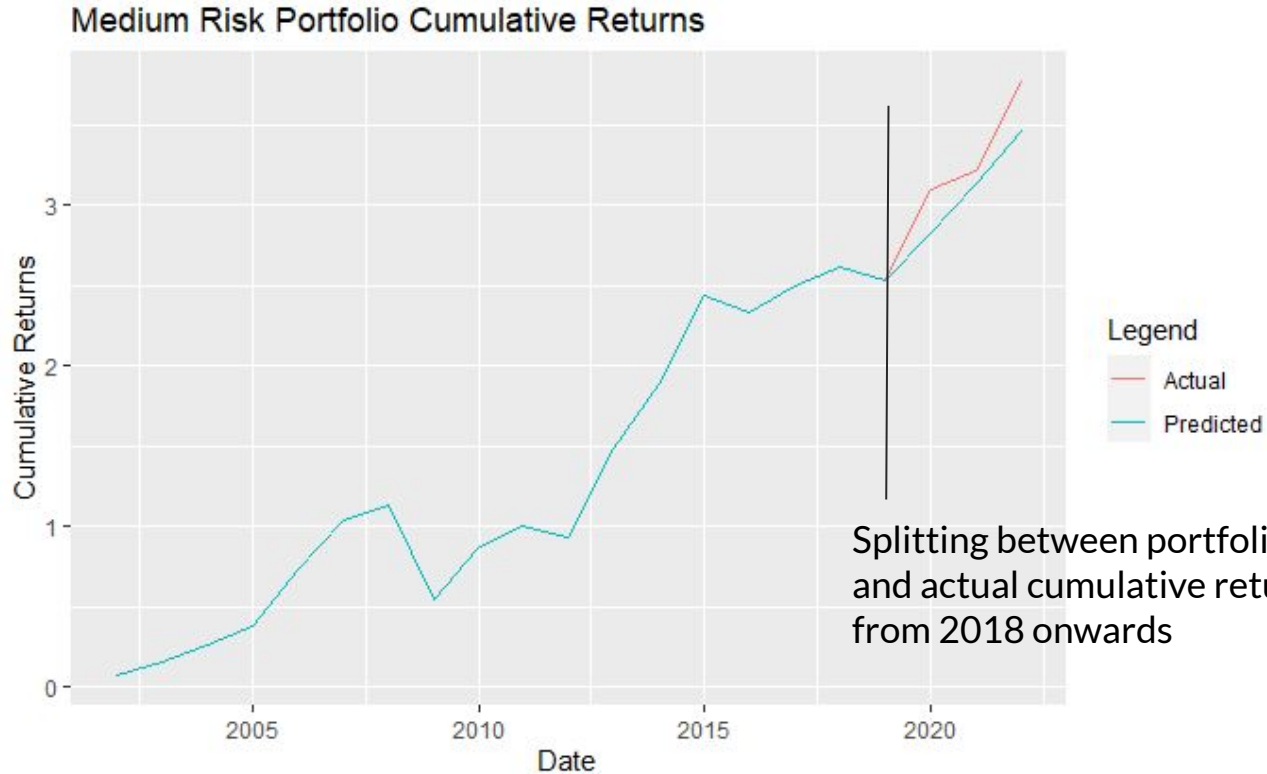
- Portfolio backtesting
  - For our data we have split it into T1 (from 2001 - 2018) and T2 (from 2018 - 2021). Using our portfolio weights we ran it with our in-sample data T1.
  - Thereafter, using the expected returns as of 2018, we forecasted it for the next three years (2018 - 2021) to obtain predicted cumulative returns.
  - By running our weights against the out-of-sample data T2, we would then obtain the actual cumulative returns from 2018 - 2021.
  - Hence, by comparing the cumulative returns from the actual and the predicted, we are able to determine the performance of our portfolio models.
  - Judging from the performance of our portfolio models, we would then rebalance our portfolios against the in-sample data T1 and repeat the backtesting process.



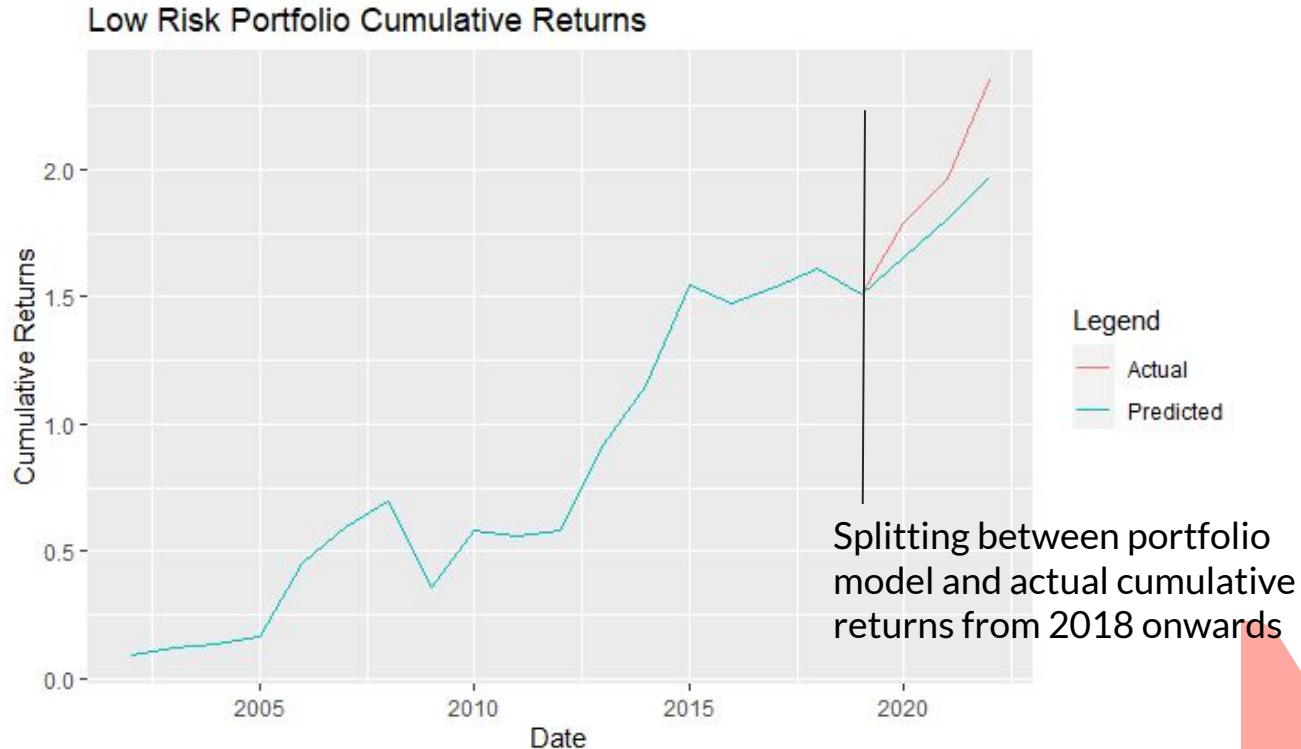
# Backtesting of High Risk Portfolio



# Backtesting of Medium Risk Portfolio



# Backtesting of Low Risk Portfolio



# Thank you

