

Carteras de inversión.

Una introducción al diseño y análisis.

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	Fundamental	Intermedio	Especializado
Finanzas	✗	✓	✗
Estadística	✓	✗	✗
R	✗	✓	✗

1 Introducción.

- Conceptos fundamentales del análisis riesgo–rendimiento.
- Construir y evaluar portafolios usando datos reales.
- Explicar la frontera media–varianza, destacando cómo se obtiene mediante métodos analíticos y numéricos.
- Cómo los pesos óptimos cambian cuando se permiten o restringen las ventas en corto.

2 Diez empresas públicas de Estados Unidos.

Ticker	Nombre de la empresa	Industria
AMD	Advanced Micro Devices, Inc.	Computación de alto rendimiento
CNC	Centene Corp.	Servicios de salud
GIS	General Mills, Inc.	Productos de consumo envasados
LMT	Lockheed Martin Corp.	Aeroespacial y defensa
LRCX	Lam Research Corp.	Semiconductores
NEM	Newmont Corp.	Minería de metales preciosos
SNPS	Synopsys, Inc.	Software de diseño
SYF	Synchrony Financial	Servicios financieros
TRMB	Trimble Inc.	Tecnología geoespacial
TTD	The Trade Desk, Inc.	Tecnología publicitaria

3 Paquetes.

```
1 library(tidyquant)
2 library(tidyverse)
3 library(lubridate)
4 library(scales)
5 library(quadprog)
```

4 Inicialización

```
1 tickers <- c("AMD", "CNC", "GIS", "LMT", "LRCX", "NEM", "SNPS", "SYF", "TRMB", "TTD")
2 n_months <- 60L
3 price_start <- ymd("2020-09-01")
4 price_end <- price_start %m+% months(n_months + 1) - days(1)
5 returns_start <- price_start %m+% months(1)
6 returns_end <- returns_start %m+% months(n_months) - months(1)
7
8 prices <- tq_get(tickers, from = price_start, to = price_end, get = "stock.prices")
9
10 monthly_returns <- prices |>
11   arrange(symbol, date) |>
12   group_by(symbol) |>
13   tq_transmute(select = adjusted, mutate_fun = periodReturn,
14                 period = "monthly", type = "arithmetic",
15                 col_rename = "monthly_return") |>
16   filter(between(date, returns_start, returns_end)) |>
17   slice_head(n = n_months) |>
18   ungroup()
19
20 stopifnot(n_distinct(monthly_returns$symbol) == length(tickers))
21
22 asset_stats <- function(data) {
23   data |>
24     group_by(symbol) |>
25     summarise(ER = mean(monthly_return, na.rm = TRUE),
26               SD = sd(monthly_return, na.rm = TRUE),
27               SR = ER / SD,
28               .groups = "drop")}
29
30 cum_index <- function(data, start_date) {
31   base <- tibble(date = start_date, symbol = unique(data$symbol), index_level = 1)
32   data |>
33     arrange(symbol, date) |>
34     group_by(symbol) |>
35     mutate(index_level = cumprod(1 + monthly_return)) |>
36     ungroup() |>
37     select(date, symbol, index_level) |>
38     bind_rows(base) |>
39     arrange(symbol, date)}
40
41 portfolio_series <- function(data, weights, label, start_date) {
42   data |>
43     filter(symbol %in% names(weights)) |>
44     group_by(date) |>
45     summarise(portfolio_return = sum(monthly_return * weights[symbol]), .groups = "drop") |>
46     arrange(date) |>
47     mutate(index_level = cumprod(1 + portfolio_return), symbol = label) |>
48     select(date, symbol, index_level) |>
49     bind_rows(tibble(date = start_date, symbol = label, index_level = 1))}
50
51 portfolio_stats <- function(data, weights, label) {
52   pr <- data |>
53     filter(symbol %in% names(weights)) |>
54     group_by(date) |>
55     summarise(portfolio_return = sum(monthly_return * weights[symbol]), .groups = "drop")
```

```

56
57 tibble(symbol = label,
58         ER = mean(pr$portfolio_return, na.rm = TRUE),
59         SD = sd(pr$portfolio_return, na.rm = TRUE),
60         SR = ER / SD)}
61
62 returns_wide <- monthly_returns |>
63   select(date, symbol, monthly_return) |>
64   pivot_wider(names_from = symbol, values_from = monthly_return) |>
65   arrange(date)
66
67 returns_mat_all <- returns_wide |>
68   select(-date) |>
69   as.matrix()
70
71 mu_vec_all <- colMeans(returns_mat_all, na.rm = TRUE)
72 cov_mat_all <- cov(returns_mat_all, use = "pairwise.complete.obs")
73 n_assets <- length(mu_vec_all)
74
75 stats_all <- tibble(symbol = names(mu_vec_all),
76                       ER = mu_vec_all,
77                       SD = apply(returns_mat_all, 2, sd, na.rm = TRUE))
78
79 color_values_all <- setNames(scales::hue_pal()(length(mu_vec_all)),
80                               names(mu_vec_all))

```

5 Matriz de correlaciones.

```
1 corr_mat <- returns_wide |>
2   select(-date) |>
3   cor(use = "pairwise.complete.obs")
4
5 print(round(corr_mat, 2), quote = FALSE, right = TRUE)
```

	##	AMD	CNC	GIS	LMT	LRCX	NEM	SNPS	SYF	TRMB	TTD
## AMD	1.00	-0.10	-0.16	-0.12	0.71	-0.06	0.70	0.35	0.50	0.51	
## CNC	-0.10	1.00	0.37	0.37	0.14	0.20	-0.16	0.12	0.01	-0.18	
## GIS	-0.16	0.37	1.00	0.46	-0.20	0.25	-0.27	-0.15	-0.20	-0.20	
## LMT	-0.12	0.37	0.46	1.00	0.00	0.32	-0.22	0.20	0.01	-0.10	
## LRCX	0.71	0.14	-0.20	0.00	1.00	0.04	0.56	0.59	0.61	0.38	
## NEM	-0.06	0.20	0.25	0.32	0.04	1.00	-0.04	-0.08	-0.05	-0.26	
## SNPS	0.70	-0.16	-0.27	-0.22	0.56	-0.04	1.00	0.33	0.48	0.31	
## SYF	0.35	0.12	-0.15	0.20	0.59	-0.08	0.33	1.00	0.67	0.29	
## TRMB	0.50	0.01	-0.20	0.01	0.61	-0.05	0.48	0.67	1.00	0.44	
## TTD	0.51	-0.18	-0.20	-0.10	0.38	-0.26	0.31	0.29	0.44	1.00	

6 Pares de correlaciones mínimas.

```
1 lowest_pairs <- corr_mat |>
2   as_tibble(rownames = "symbol_i") |>
3   pivot_longer(-symbol_i, names_to = "symbol_j", values_to = "corr") |>
4   filter(symbol_i < symbol_j) |>
5   arrange(corr) |>
6   slice_head(n = 10)
7
8 print(lowest_pairs)
```

```
## # A tibble: 10 x 3
##       symbol_i symbol_j     corr
##       <chr>     <chr>     <dbl>
## 1   GIS        SNPS    -0.266
## 2   NEM        TTD     -0.263
## 3   LMT        SNPS    -0.217
## 4   GIS        TTD     -0.201
## 5   GIS        LRCX    -0.201
## 6   GIS        TRMB    -0.198
## 7   CNC        TTD     -0.180
## 8   AMD        GIS     -0.162
## 9   CNC        SNPS    -0.156
## 10  GIS        SYF     -0.148
```

7 Riesgo-rendimiento: 10 activos individuales.

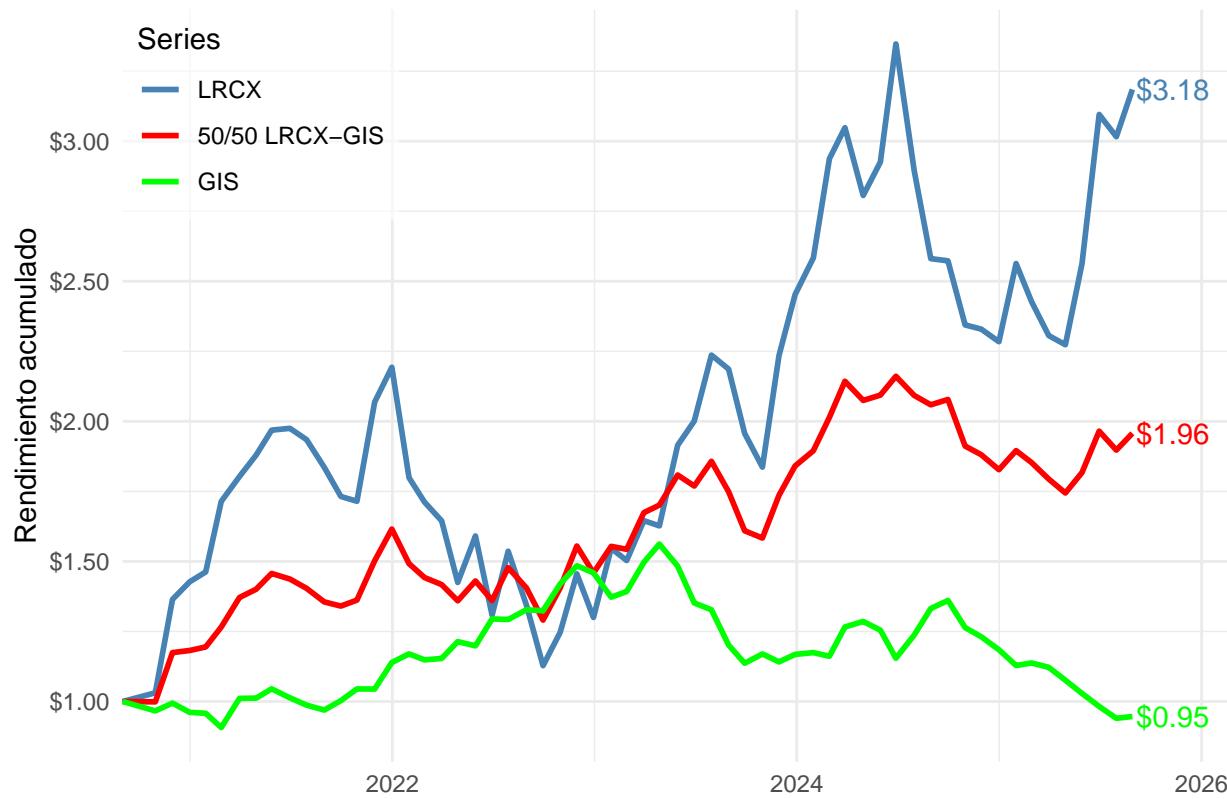
```
1 stats <- asset_stats(monthly_returns)
2
3 print(stats)

## # A tibble: 10 x 4
##   symbol      ER      SD      SR
##   <chr>     <dbl>    <dbl>    <dbl>
## 1 AMD      0.0223   0.150   0.149
## 2 CNC     -0.00542  0.102  -0.0532
## 3 GIS      0.000182  0.0477  0.00382
## 4 LMT      0.00709  0.0647  0.110
## 5 LRCX     0.0257   0.111   0.231
## 6 NEM      0.0107   0.103   0.103
## 7 SNPS     0.0216   0.0902  0.240
## 8 SYF      0.0259   0.106   0.244
## 9 TRMB     0.0132   0.0970  0.136
## 10 TTD     0.0170   0.184   0.0926
```

8 Rendimiento acumulado, cartera ingenua.

```
1 focus <- c("LRCX", "GIS")
2 color_values <- c("LRCX"="steelblue", "GIS"="green", "50/50 LRCX-GIS"="red")
3
4 cum_focus <- monthly_returns |>
5   filter(symbol %in% focus) |>
6   cum_index(price_start)
7
8 portfolio_50_50 <- portfolio_series(monthly_returns,
9                                         c(LRCX = 0.5, GIS = 0.5),
10                                        "50/50 LRCX-GIS", price_start)
11
12 plot_data <- bind_rows(cum_focus, portfolio_50_50)
13
14 last_vals <- plot_data |>
15   group_by(symbol) |>
16   summarise(final_value = last(index_level), .groups = "drop") |>
17   arrange(desc(final_value))
18
19 label_data <- plot_data |>
20   group_by(symbol) |>
21   filter(date == max(date)) |>
22   mutate(label = paste0("$", formatC(index_level, format="f", digits=2))) |>
23   ungroup()
24
25 ggplot(plot_data, aes(date, index_level, color = factor(symbol, levels = last_vals$symbol))) +
26   geom_line(linewidth = 1) +
27   geom_text(data = label_data, aes(label = label), hjust = -0.05, vjust = 0.5, show.legend =
28     FALSE) +
29   scale_color_manual(values = color_values, limits = last_vals$symbol) +
30   scale_x_date(expand = expansion(mult = c(0, 0.1))) +
31   scale_y_continuous(labels = dollar_format(prefix = "$")) +
32   labs(title = "Rendimiento acumulado.", y = "Rendimiento acumulado", x = NULL, color =
33     "Series") +
34   theme_minimal() +
35   theme(legend.position = c(0,1), legend.justification = c(0,1),
36         legend.background = element_rect(fill = alpha("white", 0.6), color = NA))
```

Rendimiento acumulado.



9 Riesgo-rendimiento: cartera ingenua y componentes.

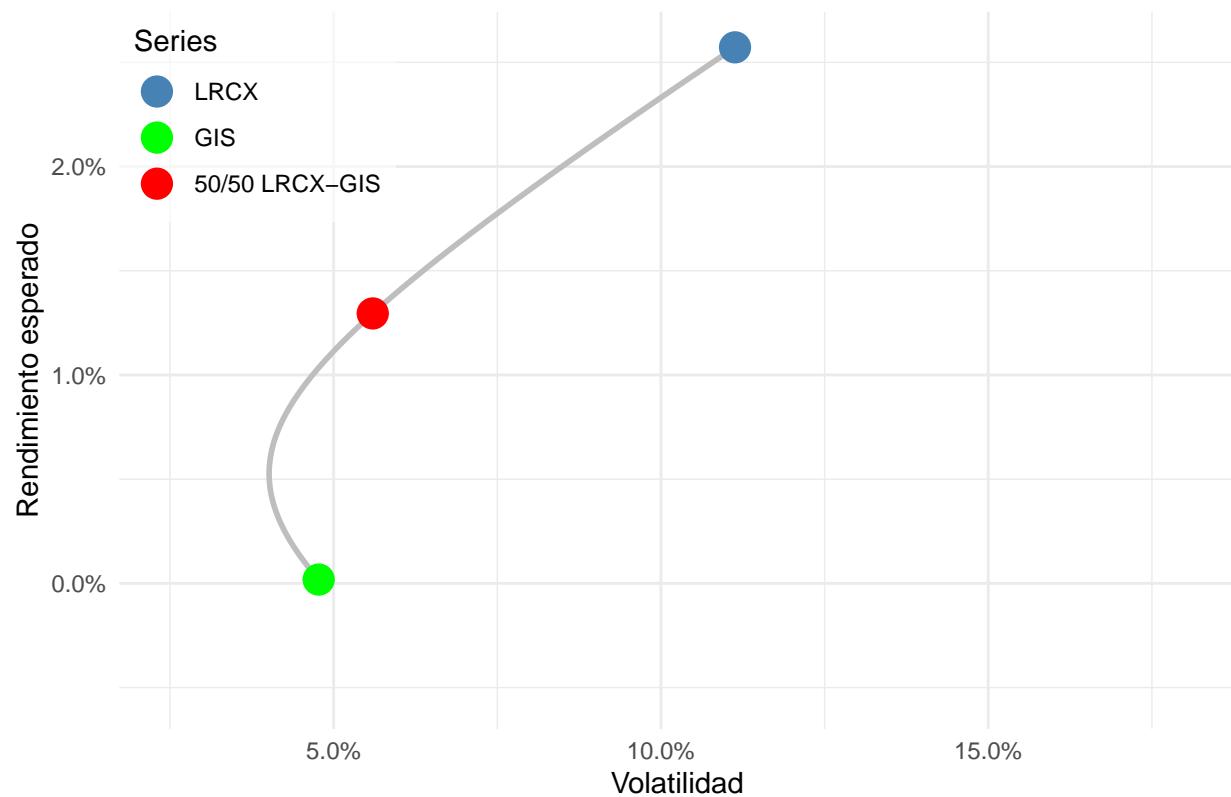
```
1 weights_50_50 <- c(LRCX = 0.5, GIS = 0.5)
2 stats_lrcx_gis <- stats |> filter(symbol %in% names(weights_50_50))
3 portfolio_stats_50_50 <- portfolio_stats(monthly_returns, weights_50_50, "50/50 LRCX-GIS")
4
5 final_table <- bind_rows(stats_lrcx_gis, portfolio_stats_50_50)
6
7 print(final_table)

## # A tibble: 3 x 4
##   symbol           ER      SD      SR
##   <chr>        <dbl>  <dbl>  <dbl>
## 1 GIS          0.000182 0.0477 0.00382
## 2 LRCX         0.0257   0.111   0.231
## 3 50/50 LRCX-GIS 0.0130  0.0560 0.231
```

10 Frontera media-varianza: 2 activos.

```
1 returns_matrix_2 <- returns_wide |>
2   select(all_of(focus))
3
4 cov_mat_2 <- cov(returns_matrix_2, use = "pairwise.complete.obs")
5 mu_vec_2 <- colMeans(returns_matrix_2, na.rm = TRUE)
6
7 frontier_data <- tibble(weight_lrcx = seq(0, 1, length.out = 200)) |>
8   mutate(weight_gis = 1 - weight_lrcx,
9         ER = weight_lrcx * mu_vec_2["LRCX"] + weight_gis * mu_vec_2["GIS"],
10        variance = weight_lrcx^2 * var(returns_matrix_2$LRCX, na.rm = TRUE) +
11          weight_gis^2 * var(returns_matrix_2$GIS, na.rm = TRUE) +
12          2 * weight_lrcx * weight_gis * cov_mat_2["LRCX", "GIS"],
13        SD = sqrt(variance))
14
15 mean_var_points <- final_table |>
16   mutate(symbol = factor(symbol, levels = c("LRCX", "GIS", "50/50 LRCX-GIS"))) |>
17   arrange(symbol)
18
19 ggplot() +
20   geom_path(data = frontier_data, aes(SD, ER), color = "grey", linewidth = 1) +
21   geom_point(data = mean_var_points, aes(SD, ER, color = symbol), size = 5) +
22   scale_color_manual(values = color_values, limits = levels(mean_var_points$symbol)) +
23   scale_x_continuous(limits = c(0.025, 0.18), labels = percent_format(accuracy = 0.1)) +
24   scale_y_continuous(limits = c(min(mu_vec_all), max(mu_vec_all)),
25                     labels = percent_format(accuracy = 0.1)) +
26   labs(title = "Frontera media-varianza.",
27        x = "Volatilidad", y = "Rendimiento esperado", color = "Series") +
28   theme_minimal() +
29   theme(legend.position = c(0,1), legend.justification = c(0,1),
30         legend.background = element_rect(fill = alpha("white", 0.6), color = NA))
```

Frontera media-varianza.



11 Cartera de mínima varianza: método numérico de optimización.

```
1 minvar2_label <- "Min-Var 2-Asset (LRCX-GIS)"
2
3 assets_2 <- focus
4 qp_2 <- solve.QP(Dmat = 2 * cov_mat_2,
5                      dvec = rep(0, length(assets_2)),
6                      Amat = matrix(1, nrow = length(assets_2), ncol = 1),
7                      bvec = 1, meq = 1)
8
9 weights_minvar2_qp <- setNames(qp_2$solution, assets_2)
10
11 print(weights_minvar2_qp)

##          LRCX          GIS
## 0.1990057 0.8009943
```

12 Cartera de mínima varianza: método analítico algebráico para 2 activos.

$$w_1^* = \frac{\sigma_2^2 - \sigma_{12}}{\sigma_1^2 + \sigma_2^2 - 2\sigma_{12}}.$$

```
1 Sigma_2 <- cov_mat_2[assets_2, assets_2]
2 var1 <- Sigma_2[1, 1]; var2 <- Sigma_2[2, 2]; cov12 <- Sigma_2[1, 2]
3 w1 <- (var2 - cov12) / (var1 + var2 - 2 * cov12)
4 weights_minvar2_closed <- setNames(c(w1, 1 - w1), assets_2)
5
6 print(weights_minvar2_closed)
```

```
##          LRCX          GIS
## 0.1990057 0.8009943
```

13 Cartera de mínima varianza: método analítico matricial para 2 activos.

$$w^* = \frac{\Sigma^{-1} \mathbf{1}}{\mathbf{1}^\top \Sigma^{-1} \mathbf{1}}$$

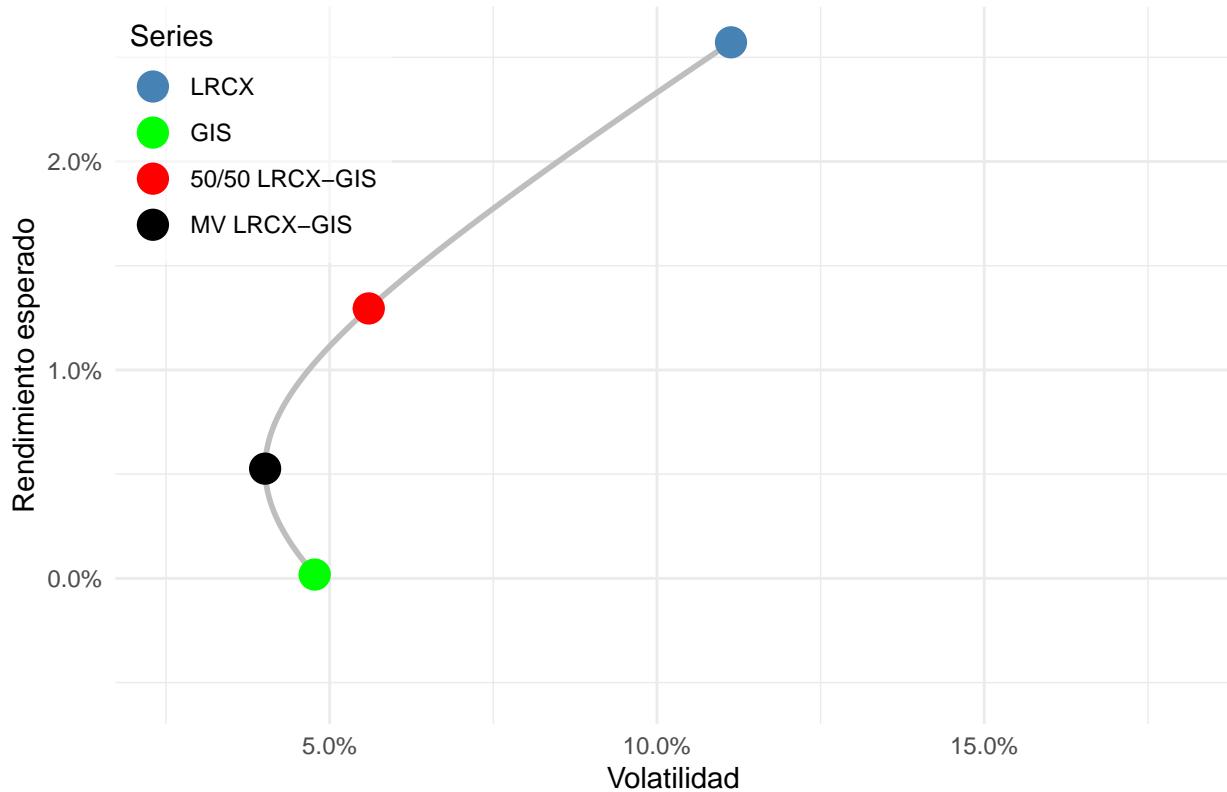
```
1 Sigma_2 <- cov_mat_2[assets_2, assets_2, drop = FALSE]
2 ones <- rep(1, length(assets_2))
3 Sigma_inv <- solve(Sigma_2)
4 num <- Sigma_inv %*% ones
5 den <- as.numeric(t(ones) %*% Sigma_inv %*% ones)
6
7 w_star <- as.numeric(num / den)
8 names(w_star) <- assets_2
9
10 print(w_star)
```

```
##          LRCX          GIS
## 0.1990057 0.8009943
```

14 Frontera media-varianza: 2 activos, cartera MV.

```
1 minvar2_label <- "MV LRCX-GIS"
2 minvar_point_2 <-
3   tibble(symbol = minvar2_label,
4         ER = sum(weights_minvar2_qp * mu_vec_2),
5         SD = sqrt(as.numeric(t(weights_minvar2_qp) %*% cov_mat_2 %*% weights_minvar2_qp)))
6
7 mean_var_points_ext <- bind_rows(stats_lrcx_gis,
8                                   portfolio_stats_50_50,
9                                   minvar_point_2) |>
10  mutate(symbol = factor(symbol,
11                  levels = c("LRCX", "GIS", "50/50 LRCX-GIS", minvar2_label))) |>
12  arrange(symbol)
13
14 color_values_ext <- c(color_values, setNames("black", minvar2_label))
15
16 ggplot() +
17   geom_path(data = frontier_data, aes(SD, ER), color = "grey", linewidth = 1) +
18   geom_point(data = mean_var_points_ext, aes(SD, ER, color = symbol), size = 5) +
19   scale_color_manual(values = color_values_ext, limits = levels(mean_var_points_ext$symbol)) +
20   scale_x_continuous(limits = c(0.025, 0.18), labels = percent_format(accuracy = 0.1)) +
21   scale_y_continuous(limits = c(min(mu_vec_all), max(mu_vec_all)), labels =
22     ↪ percent_format(accuracy = 0.1)) +
23   labs(title = "Frontera media-varianza.",
24        x = "Volatilidad", y = "Rendimiento esperado", color = "Series") +
25   theme_minimal() +
26   theme(legend.position = c(0, 1),
27         legend.justification = c(0, 1),
28         legend.background = element_rect(fill = alpha("white", 0.6), color = NA))
```

Frontera media-varianza.



15 Cartera de mínima varianza: método numérico para 10 activos.

Pesos positivos.

```
1 minvar10_ns_label <- "MV (no shorts)"
2
3 weights_minvar10_ns <-
4   solve.QP(2 * cov_mat_all, rep(0, length(mu_vec_all)),
5           cbind(rep(1, length(mu_vec_all)), diag(length(mu_vec_all))),
6           c(1, rep(0, length(mu_vec_all))), meq = 1)$solution
7
8 names(weights_minvar10_ns) <- names(mu_vec_all)
9
10 minvar_point_10_ns <-
11   tibble(symbol = minvar10_ns_label,
12         ER = sum(weights_minvar10_ns * mu_vec_all),
13         SD = sqrt(as.numeric(t(weights_minvar10_ns) %*% cov_mat_all %*%
14                           weights_minvar10_ns)),
15         SR = ER/SD)
16 minvar_point_10_ns

## # A tibble: 1 x 4
##   symbol             ER      SD      SR
##   <chr>        <dbl>  <dbl>  <dbl>
## 1 MV (no shorts) 0.00717 0.0337 0.213
```

16 Cartera de mínima varianza: método numérico para 10 activos, pesos.

```
1 minvar_weights_table_ns <-
2   tibble(symbol = names(mu_vec_all),
3         weight = weights_minvar10_ns,
4         weight_pct = scales::percent(weights_minvar10_ns, accuracy = 0.01)) |>
5   arrange(desc(weight)) |>
6   bind_rows(tibble(symbol = "Total",
7                     weight = sum(weights_minvar10_ns),
8                     weight_pct = scales::percent(sum(weights_minvar10_ns),
9                                         accuracy = 0.01)))
10
11 print(minvar_weights_table_ns)
```



```
## # A tibble: 11 x 3
##       symbol     weight weight_pct
##       <chr>      <dbl>    <chr>
## 1   GIS      5.40e- 1 54.01%
## 2 SNPS     1.83e- 1 18.34%
## 3 LMT      1.12e- 1 11.17%
## 4 TRMB     4.62e- 2 4.62%
## 5 NEM      4.52e- 2 4.52%
## 6 SYF      3.05e- 2 3.05%
## 7 TTD      3.00e- 2 3.00%
## 8 CNC      1.30e- 2 1.30%
## 9 AMD      1.34e-18 0.00%
## 10 LRCX     0          0.00%
## 11 Total    1          e+ 0 100.00%
```

17 Cartera de mínima varianza: método analítico matricial para 10 activos, pesos.

$w^* = \frac{\Sigma^{-1}\mathbf{1}}{\mathbf{1}^\top \Sigma^{-1}\mathbf{1}}$. Markowitz, ventas en corto permitidas.

```

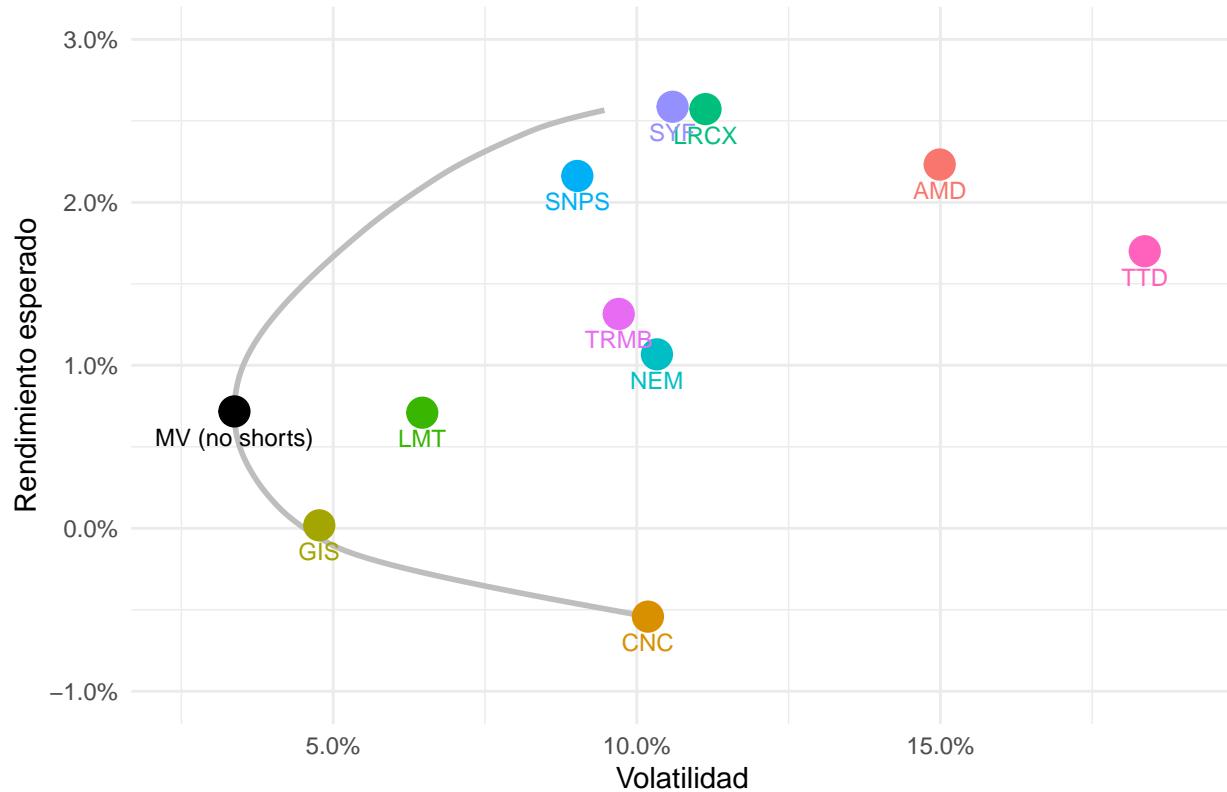
1 minvar10_short_label <- "MV (shorts)"
2
3 Sigma_all <- cov_mat_all
4 ones_all <- rep(1, length(mu_vec_all))
5 inv_Sigma_all <- solve(Sigma_all)
6 A_all <- as.numeric(t(ones_all) %*% inv_Sigma_all %*% ones_all)
7
8 weights_minvar10_short <- as.numeric(inv_Sigma_all %*% ones_all / A_all)
9 names(weights_minvar10_short) <- names(mu_vec_all)
10
11 minvar_point_10_short <-
12   tibble(symbol = minvar10_short_label, ER = sum(weights_minvar10_short * mu_vec_all),
13         SD = sqrt(as.numeric(t(weights_minvar10_short) %*% Sigma_all %*%
14             weights_minvar10_short)))
15
16 weights_gmv_table_short <-
17   tibble(symbol = names(mu_vec_all),
18         weight = weights_minvar10_short,
19         weight_pct = scales::percent(weights_minvar10_short, accuracy = 0.01)) |>
20   arrange(desc(weight)) |>
21   bind_rows(tibble(symbol = "Total",
22                 weight = sum(weights_minvar10_short),
23                 weight_pct = scales::percent(sum(weights_minvar10_short), accuracy =
24                   0.01)))
25
26 print(weights_gmv_table_short)

## # A tibble: 11 x 3
##       symbol     weight weight_pct
##       <chr>      <dbl> <chr>
## 1  GIS        0.539  53.93%
## 2 SNPS       0.240  24.04%
## 3 LMT        0.106  10.62%
## 4 TRMB       0.0490 4.90%
## 5 TTD        0.0469 4.69%
## 6 NEM        0.0419 4.19%
## 7 LRCX       0.0416 4.16%
## 8 SYF        0.0187 1.87%
## 9 CNC        0.00510 0.51%
## 10 AMD       -0.0893 -8.93%
## 11 Total      1        100.00%
```

18 Frontera media-varianza: 10 activos, sin ventas en corto.

```
1 target_returns <- seq(min(mu_vec_all), max(mu_vec_all), length.out = 150)
2 Amat_ns <- rbind(rep(1, n_assets), mu_vec_all, diag(n_assets))
3
4 frontier_points_ns <- map_dfr(target_returns, function(tr) {
5   bvec <- c(1, tr, rep(0, n_assets))
6   sol <- tryCatch(solve.QP(2 * cov_mat_all, rep(0, n_assets),
7                     t(Amat_ns), bvec, meq = 2),
8                     error = function(e) NULL)
9   if (is.null(sol)) return(NULL)
10  tibble(ER = tr,
11         SD = sqrt(as.numeric(t(sol$solution) %*% cov_mat_all %*% sol$solution))))} |>
12  drop_na()
13
14 stats_all_ext_ns <- bind_rows(stats_all, minvar_point_10_ns)
15 color_values_ext_ns <- c(color_values_all, setNames("black", minvar10_ns_label))
16
17 ggplot() +
18   geom_path(data = frontier_points_ns, aes(SD, ER), color = "grey", linewidth = 1) +
19   geom_point(data = stats_all_ext_ns, aes(SD, ER, color = symbol), size = 5) +
20   geom_text(data = stats_all_ext_ns, aes(SD, ER, label = symbol, color = symbol),
21             vjust = 2, size = 3, show.legend = FALSE) +
22   scale_color_manual(values = color_values_ext_ns, limits = stats_all_ext_ns$symbol) +
23   scale_x_continuous(limits = c(0.025, 0.19), labels = percent_format(accuracy = 0.1)) +
24   scale_y_continuous(limits = c(-0.01, 0.03), labels = percent_format(accuracy = 0.1)) +
25   labs(title = "Frontera media-varianza, sin ventas en corto.",
26        x = "Volatilidad", y = "Rendimiento esperado") +
27   guides(color = "none") +
28   theme_minimal()
```

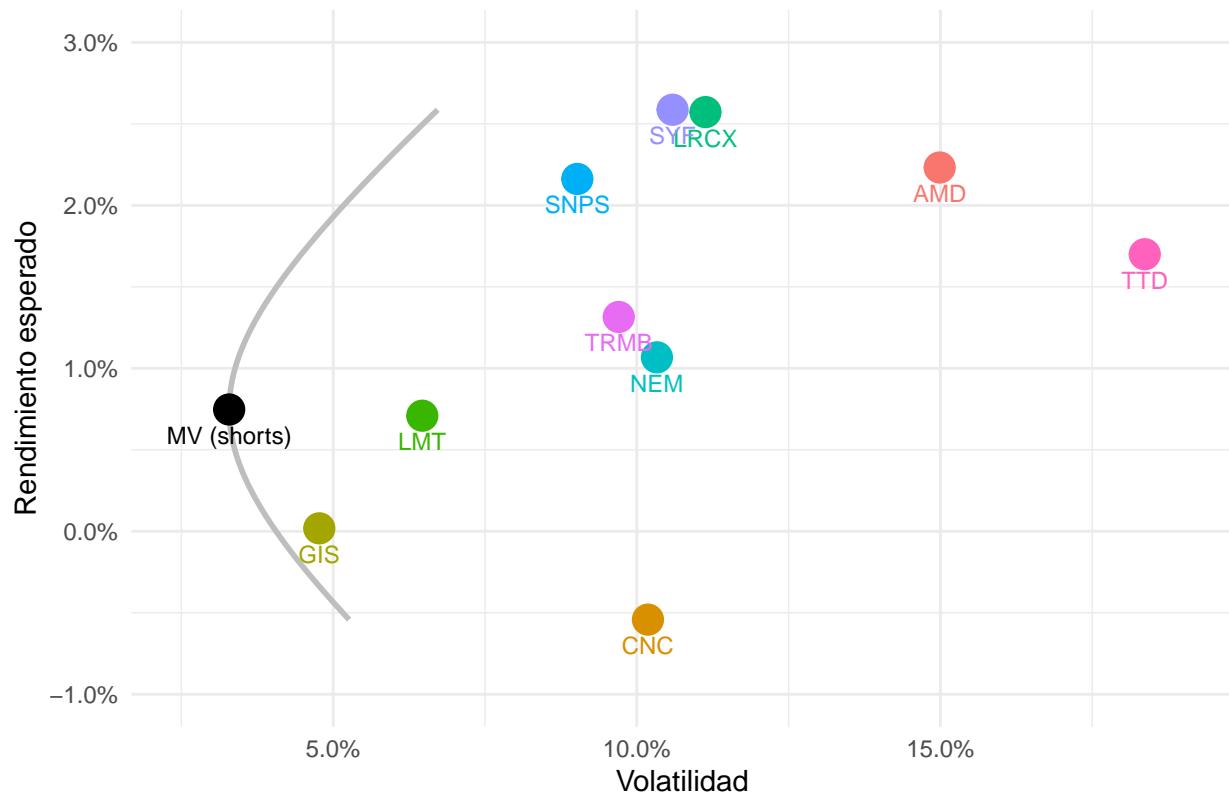
Frontera media–varianza, sin ventas en corto.



19 Frontera media-varianza: 10 activos, con ventas en corto.

```
1 Amat_short <- rbind(rep(1, n_assets), mu_vec_all)
2
3 frontier_points_short <- map_dfr(target_returns, function(tr) {
4   bvec <- c(1, tr)
5   sol <- tryCatch(solve.QP(2 * cov_mat_all, rep(0, n_assets),
6                     t(Amat_short), bvec, meq = 2),
7                     error = function(e) NULL)
8   if (is.null(sol)) return(NULL)
9   tibble(ER = tr,
10        SD = sqrt(as.numeric(t(sol$solution) %*% cov_mat_all %*% sol$solution)))) |>
11 drop_na()
12
13 stats_all_ext_short <- bind_rows(stats_all, minvar_point_10_short)
14 color_values_ext_short <- c(color_values_all, setNames("black", minvar10_short_label))
15
16 ggplot() +
17   geom_path(data = frontier_points_short, aes(SD, ER), color = "grey", linewidth = 1) +
18   geom_point(data = stats_all_ext_short, aes(SD, ER, color = symbol), size = 5) +
19   geom_text(data = stats_all_ext_short, aes(SD, ER, label = symbol, color = symbol),
20             vjust = 2, size = 3, show.legend = FALSE) +
21   scale_color_manual(values = color_values_ext_short, limits = stats_all_ext_short$symbol) +
22   scale_x_continuous(limits = c(0.025, 0.19), labels = percent_format(accuracy = 0.1)) +
23   scale_y_continuous(limits = c(-0.01, 0.03),
24                      labels = percent_format(accuracy = 0.1)) +
25   labs(title = "Fronteria media-varianza, con ventas en corto.",
26        x = "Volatilidad", y = "Rendimiento esperado") +
27   guides(color = "none") +
28   theme_minimal()
```

Fronteria media–varianza, con ventas en corto.



20 Riesgo-rendimiento: 10 activos individuales y carteras.

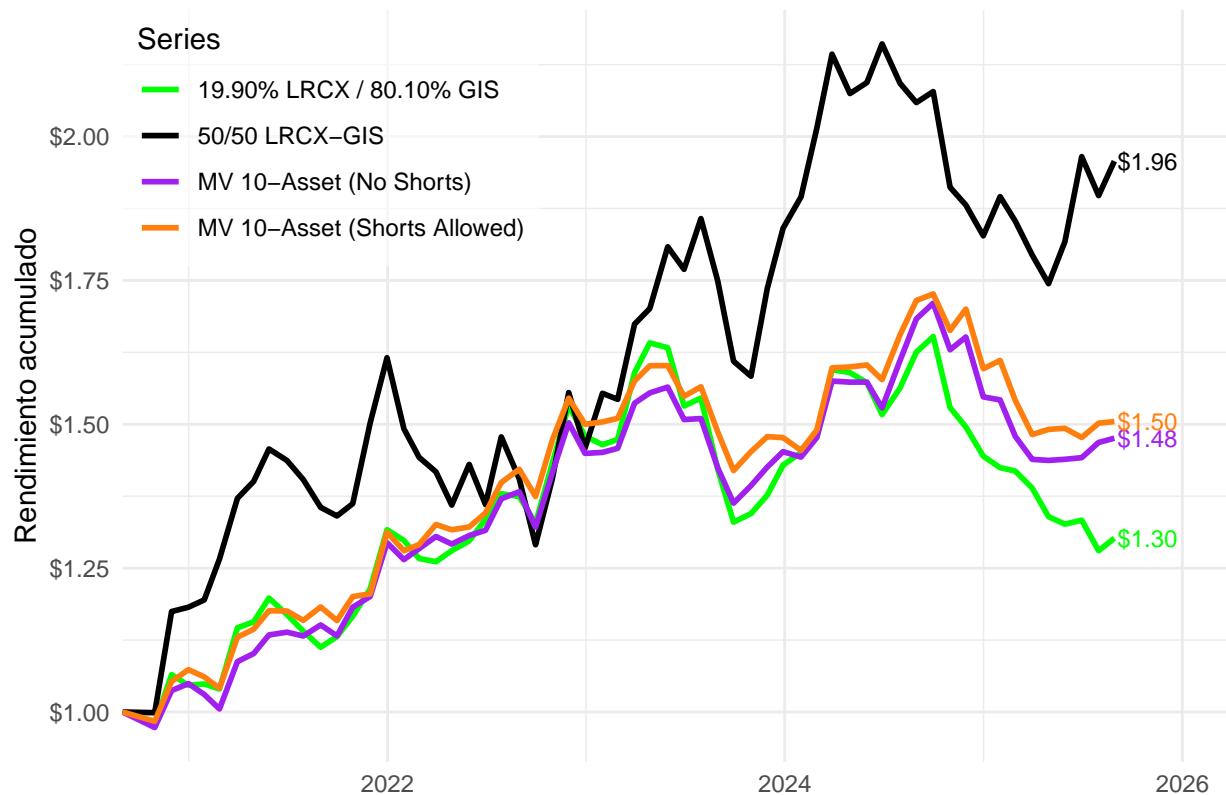
```
1 weights_50_50 <- c(LRCX = 0.5, GIS = 0.5)
2 custom_weights <- c(LRCX = 0.1990055, GIS = 0.8009945)
3 portfolio_label <- "19.90% LRCX / 80.10% GIS"
4
5 weights_specs <-
6   list("50/50 LRCX-GIS" = weights_50_50,
7     "19.90% LRCX / 80.10% GIS" = custom_weights,
8     "MV 10-Asset (No Shorts)" = weights_minvar10_ns,
9     "MV 10-Asset (Shorts Allowed)" = weights_minvar10_short)
10
11 portfolio_stats_ext <- imap_dfr(weights_specs, ~ portfolio_stats(monthly_returns, .x, .y))
12
13 final_table <- bind_rows(asset_stats(monthly_returns), portfolio_stats_ext) |>
14   mutate(symbol = factor(symbol, levels = c(tickers, names(weights_specs)))) |>
15   arrange(symbol)
16
17 print(final_table)

## # A tibble: 14 x 4
##       symbol               ER      SD      SR
##       <fct>        <dbl>    <dbl>    <dbl>
## 1 AMD            0.0223   0.150   0.149
## 2 CNC           -0.00542  0.102  -0.0532
## 3 GIS            0.000182  0.0477  0.00382
## 4 LMT            0.00709  0.0647  0.110
## 5 LRCX           0.0257   0.111   0.231
## 6 NEM            0.0107   0.103   0.103
## 7 SNPS           0.0216   0.0902  0.240
## 8 SYF            0.0259   0.106   0.244
## 9 TRMB           0.0132   0.0970  0.136
## 10 TTD            0.0170   0.184   0.0926
## 11 50/50 LRCX-GIS 0.0130   0.0560  0.231
## 12 19.90% LRCX / 80.10% GIS 0.00526  0.0401  0.131
## 13 MV 10-Asset (No Shorts) 0.00717  0.0337  0.213
## 14 MV 10-Asset (Shorts Allowed) 0.00747  0.0329  0.227
```

21 Rendimiento acumulado, todas las carteras.

```
1 portfolio_cum <- imap_dfr(weights_specs, ~ portfolio_series(monthly_returns, .x, .y,
2   ~ price_start))
3 last_values_port <- portfolio_cum |>
4   group_by(symbol) |>
5   summarise(final_value = last(index_level), .groups = "drop") |>
6   arrange(desc(final_value))
7
8 series_levels <- last_values_port$symbol
9
10 palette_base <- c("50/50 LRCX-GIS" = "black",
11                      "19.90% LRCX / 80.10% GIS" = "green",
12                      "MV 10-Asset (No Shorts)" = "purple",
13                      "MV 10-Asset (Shorts Allowed)" = "#ff7f0e")
14
15 portfolio_colors <- palette_base[series_levels]
16
17 portfolio_cum <- portfolio_cum |>
18   mutate(symbol = factor(symbol, levels = series_levels))
19
20 label_data_port <- portfolio_cum |>
21   group_by(symbol) |> filter(date == max(date)) |>
22   mutate(label = paste0("$", formatC(index_level, format="f", digits=2))) |>
23   ungroup()
24
25 ggplot(portfolio_cum, aes(date, index_level, color = symbol)) +
26   geom_line(linewidth = 1) +
27   geom_text(data = label_data_port, aes(label = label),
28             hjust = -0.05, vjust = 0.5, size = 3.2, show.legend = FALSE) +
29   scale_color_manual(values = portfolio_colors, breaks = series_levels,
30                      limits = series_levels, drop = FALSE) +
31   scale_x_date(expand = expansion(mult = c(0, 0.12))) +
32   scale_y_continuous(labels = dollar_format(prefix = "$")) +
33   labs(title = "Rendimiento acumulado.",
34        y = "Rendimiento acumulado", x = NULL, color = "Series") +
35   theme_minimal() +
36   theme(legend.position = c(0, 1), legend.justification = c(0, 1),
37         legend.background = element_rect(fill = alpha("white", 0.6), color = NA))
```

Rendimiento acumulado.



22 Conclusión.

- La diversificación reduce riesgo y permite construir portafolios con mejor relación riesgo–rendimiento que los activos individuales.
- La frontera media–varianza es una herramienta poderosa para visualizar las decisiones de inversión y comprender el intercambio entre riesgo y retorno.