

# Elementary Effects for the BTD Model

## Setup packages.

```
In [1]: require(data.table)
require(magrittr)
require(sensitivity)

require(ggplot2)
```

```
Loading required package: data.table
Loading required package: magrittr
Loading required package: sensitivity
Registered S3 method overwritten by 'sensitivity':
  method      from
  print.src dplyr
Loading required package: ggplot2
```

## Design experiment.

### Load input ranges.

```
In [ ]: z.ranges <- fread("input-ranges.tsv")
z.ranges %>% dim
```

```
In [ ]: z.ranges %>% summary
```

### One-at a time experiment with 500 repetitions, à la Morris.

```
In [ ]: z.design <- morris(
  NULL,
  factors = z.ranges$Variable,
  r = 500,
  design = list(
    type = "oat",
    levels = mapply(function(t, x0, x1) {
      if (t == "Integer")
        x1 - x0 + 1
      else if (t == "Boolean")
        2
      else
        5
    }, z.ranges$Type, z.ranges$`Sensitivity Minimum`, z.ranges$`Sensitivity Maximum`),
    grid.jump = 1
  )
  z.design$X %>% dim
```

```
In [ ]: write.table(z.design$X, file = "design.tsv", row.names = FALSE, col.names = TRUE, sep = "\t", quote = FALSE)
```

## Relate the design to the model's variables.

```
In [ ]: z.inputs <- cbind(
  Run = 1:(dim(z.design$X)[1]),
  data.table(
    sweep(
      sweep(z.design$X, MARGIN = 2, z.ranges$`Sensitivity Maximum` - z.ranges$`Sensitivity Minimum`, `*`),
      MARGIN = 2,
      z.ranges$`Sensitivity Minimum`,
      `+`
    )
  )
)
z.inputs %>% summary
```

```
In [ ]: write.table(z.inputs, file="inputs.tsv", row.names = FALSE, col.names = TRUE, sep = "\t", quote = FALSE)
```

## Analyze results.

### Read files.

### Read design.

```
In [2]: z.design <- fread("design.tsv")  
z.design %>% dim
```

```
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```

### Read inputs.

```
In [3]: z.inputs <- fread("inputs.tsv")  
z.inputs %>% dim
```

```
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```

### Read outputs.

```
In [4]: z.outputs <- fread("outputs.tsv")  
z.outputs[Time == 2050] %>% dim
```

```
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```

```
In [5]: z.outputs[Time == 2050] %>% summary
```

Run	Time	bioproduct market share	mass
Min. : 1	Min. :2050	Min. : -27.65680	
1st Qu.:10626	1st Qu.:2050	1st Qu.: 0.01149	
Median :21250	Median :2050	Median : 0.33107	
Mean :21250	Mean :2050	Mean : 0.59533	
3rd Qu.:31875	3rd Qu.:2050	3rd Qu.: 0.71766	
Max. :42500	Max. :2050	Max. : 54.99460	
current market size economic current market size mass long term market share			
Min. : -9.881e+13	Min. :1.750e+04	Min. :0.0000	
1st Qu.: -2.500e+10	1st Qu.:2.069e+07	1st Qu.:0.2126	
Median : 7.524e+10	Median :6.609e+07	Median :0.3172	
Mean : 3.949e+11	Mean :2.866e+08	Mean :0.3959	
3rd Qu.: 5.717e+11	3rd Qu.:2.009e+08	3rd Qu.:0.5406	
Max. : 5.844e+13	Max. :7.965e+09	Max. :0.9810	
long term market value	Adopters	NonAdopters	
Min. :0.000e+00	Min. : -11334900	Min. :5.130e+02	
1st Qu.:0.000e+00	1st Qu.: 416403	1st Qu.:5.613e+06	
Median :0.000e+00	Median : 7035080	Median :2.763e+07	
Mean :2.098e+09	Mean : 135755526	Mean :1.205e+08	
3rd Qu.:7.470e+02	3rd Qu.: 59639400	3rd Qu.:9.357e+07	
Max. :6.297e+11	Max. :5976570000	Max. :5.388e+09	
Potential Adopters	abandoning bioproduct	Cumulative Demoing Producti on	
Min. : -20212000	Min. :0	Min. : 0	
1st Qu.: 20273	1st Qu.:0	1st Qu.: 0	
Median : 1188780	Median :0	Median : 0	
Mean : 30356549	Mean :0	Mean : 182732	
3rd Qu.: 19697400	3rd Qu.:0	3rd Qu.: 0	
Max. :2077400000	Max. :0	Max. :11109600	
Cumulative Production	prepiloting	pilot plant construction	
Min. : 0	Min. :0.0000	Min. :0.0000000	
1st Qu.: 0	1st Qu.:0.0000	1st Qu.:0.0000000	
Median : 0	Median :0.0000	Median :0.0000000	
Mean : 696871	Mean :0.2168	Mean :0.0002118	
3rd Qu.: 0	3rd Qu.:0.0000	3rd Qu.:0.0000000	
Max. :67509600	Max. :1.0000	Max. :1.0000000	
pilot plant is built	startup piloting	complete	piloting ongoing
Min. :0.000	Min. :0.0000	Min. :0.0000	
1st Qu.:1.000	1st Qu.:1.0000	1st Qu.:0.0000	
Median :1.000	Median :1.0000	Median :1.0000	
Mean :0.783	Mean :0.7767	Mean :0.5731	
3rd Qu.:1.000	3rd Qu.:1.0000	3rd Qu.:1.0000	
Max. :1.000	Max. :1.0000	Max. :1.0000	
piloting progress	piloting complete	predemoing	demo plant const ruction
Min. :0.00	Min. :0.0000	Min. :0.00000	Min. :0.000000
1st Qu.:0.00	1st Qu.:0.0000	1st Qu.:0.00000	1st Qu.:0.000000
Median :0.00	Median :0.0000	Median :0.00000	Median :0.000000
Mean :0.21	Mean :0.2099	Mean :0.06772	Mean :0.003106
3rd Qu.:0.00	3rd Qu.:0.0000	3rd Qu.:0.00000	3rd Qu.:0.000000
Max. :1.00	Max. :1.0000	Max. :1.00000	Max. :1.000000
demo plant is built regulatory process ongoing startup demoing complet ed			
Min. :0.0000	Min. :0.00	Min. :0.0000	
1st Qu.:0.0000	1st Qu.:0.00	1st Qu.:0.0000	
Median :0.0000	Median :0.00	Median :0.0000	

Mean :0.1372	Mean :0.01	Mean :0.1329
3rd Qu.:0.0000	3rd Qu.:0.00	3rd Qu.:0.0000
Max. :1.0000	Max. :1.00	Max. :1.0000
demoing ongoing	demoing progress	demoing complete
Min. :0.00000	Min. :0.00000	Min. :0.00000
1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.:0.00000
Median :0.00000	Median :0.00000	Median :0.00000
Mean :0.09219	Mean :0.05074	Mean :0.04504
3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.:0.00000
Max. :1.00000	Max. :1.00000	Max. :1.00000
precommercial	commercial plant	construction commercial plant is b
uilt		
Min. :0.000000	Min. :0.000000	Min. :0.000000
1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
Median :0.000000	Median :0.000000	Median :0.000000
Mean :0.007906	Mean :0.003294	Mean :0.03144
3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
Max. :1.000000	Max. :1.000000	Max. :1.000000
commercial plant operation technology readiness level stage in progres		
s		
Min. :0.00000	Min. :6.000	Min. : 2.000
1st Qu.:0.00000	1st Qu.:6.000	1st Qu.: 5.000
Median :0.00000	Median :6.000	Median : 6.000
Mean :0.03071	Mean :6.311	Mean : 8.823
3rd Qu.:0.00000	3rd Qu.:6.000	3rd Qu.:10.000
Max. :1.00000	Max. :9.000	Max. :58.000
BS equity	payback period	NPV at required return
Min. : -2.075e+09	Min. : 0.00	Min. : -3.664e+10
1st Qu.: -1.073e+07	1st Qu.: 8.78	1st Qu.: -8.530e+08
Median : -1.996e+06	Median : 12.02	Median : -2.630e+08
Mean : 2.478e+09	Mean : 28.52	Mean : 8.902e+08
3rd Qu.: 1.777e+08	3rd Qu.: 15.89	3rd Qu.: 8.030e+06
Max. : 4.108e+11	Max. :34389.80	Max. : 2.937e+11
profitability indicator	bioproduct favorability indicator	
Min. :0.0000	Min. :0.0000	
1st Qu.:1.0000	1st Qu.:0.0000	
Median :1.0000	Median :0.0000	
Mean :0.9214	Mean :0.2349	
3rd Qu.:1.0000	3rd Qu.:0.0000	
Max. :1.0000	Max. :1.0000	
long term selling price without green premium after market entry		
Min. : 369.7		
1st Qu.:1399.2		
Median :2078.5		
Mean :2136.7		
3rd Qu.:2768.2		
Max. :6965.5		
total approval cost	total approval time	in business indicator
Min. : -1599210	Min. : 0.019	Min. :0.0000
1st Qu.: 199833	1st Qu.: 1.301	1st Qu.:0.0000
Median : 1052730	Median : 3.086	Median :0.0000
Mean : 1636119	Mean : 3.762	Mean :0.4059
3rd Qu.: 2626350	3rd Qu.: 5.521	3rd Qu.:1.0000
Max. :12420100	Max. :20.166	Max. :1.0000
internal project cancelled indicator	investing	granting
Min. :0.0000000	Min. :0.000e+00	Min. :0.000
e+00		

1st Qu.:0.0000000	1st Qu.:0.000e+00	1st Qu.:0.000
e+00		
Median :0.0000000	Median :0.000e+00	Median :0.000
e+00		
Mean :0.0006588	Mean :2.162e+08	Mean :5.044
e+08		
3rd Qu.:0.0000000	3rd Qu.:0.000e+00	3rd Qu.:3.192
e+07		
Max. :1.0000000	Max. :2.189e+11	Max. :9.984
e+10		
Total Government Grants	Total Investment	Working Capital
Min. :0.000e+00	Min. :3.000e+06	Min. : -2.075e+09
1st Qu.:0.000e+00	1st Qu.:1.145e+08	1st Qu.: -1.073e+07
Median :1.373e+08	Median :7.370e+08	Median : -1.996e+06
Mean :6.579e+08	Mean :1.274e+09	Mean : 2.478e+09
3rd Qu.:3.661e+08	3rd Qu.:1.488e+09	3rd Qu.: 1.777e+08
Max. :3.754e+10	Max. :4.533e+10	Max. : 4.108e+11
IS production incentive		
Min. :0.000e+00		
1st Qu.:0.000e+00		
Median :0.000e+00		
Mean :1.075e+08		
3rd Qu.:0.000e+00		
Max. :1.261e+10		

## Compute elementary effects.

Just use the final year.

```
In [6]: z.outputs.clean <- z.outputs[order(Run)][`Time` == 2050, c(-1, -2)] %>%
as.matrix
z.outputs.clean %>% dim
```

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Define functions to compute elementary effects.

```

In [7]: ind.rep <- function(i, p) {
# indices of the points of the ith trajectory in the DoE
  (1 : (p + 1)) + (i - 1) * (p + 1)
}

ee.oat <- function(X, y) {
  # compute the elementary effects for a OAT design
  p <- ncol(X)
  r <- nrow(X) / (p + 1)

# if(is(y,"numeric")){
if(inherits(y, "numeric")){
  one_i_vector <- function(i){
    j <- ind.rep(i, p)
    j1 <- j[1 : p]
    j2 <- j[2 : (p + 1)]
    # return((y[j2] - y[j1]) / rowSums(X[j2,] - X[j1,]))
    return(solve(X[j2,] - X[j1,], y[j2] - y[j1]))
  }
  ee <- vapply(1:r, one_i_vector, FUN.VALUE = numeric(p))
  ee <- t(ee)
  # "ee" is now a (r times p)-matrix.
# } else if(is(y,"matrix")){
} else if(inherits(y, "matrix")){
  one_i_matrix <- function(i){
    j <- ind.rep(i, p)
    j1 <- j[1 : p]
    j2 <- j[2 : (p + 1)]
    return(solve(X[j2,] - X[j1,],
                  y[j2, , drop = FALSE] - y[j1, , drop = FALSE]))
  }
  ee <- vapply(1:r, one_i_matrix,
               FUN.VALUE = matrix(0, nrow = p, ncol = dim(y)[2]))
  # Special case handling for p == 1 and ncol(y) == 1 (in this case,
  "ee" is
  # a vector of length "r"):
  if(p == 1 && dim(y)[2] == 1){
    ee <- array(ee, dim = c(r, 1, 1))
  }
  # Transpose "ee" (an array of dimensions c(p, ncol(y), r)) to an array of
  # dimensions c(r, p, ncol(y)) (for better consistency with the standard
  # case that "class(y) == "numeric")):
  ee <- aperm(ee, perm = c(3, 1, 2))
# } else if(is(y,"array")){
} else if(inherits(y, "array")){
  one_i_array <- function(i){
    j <- ind.rep(i, p)
    j1 <- j[1 : p]
    j2 <- j[2 : (p + 1)]
    ee_per_3rd_dim <- sapply(1:(dim(y)[3]), function(idx_3rd_dim){
      y_j2_matrix <- y[j2, , idx_3rd_dim]
      y_j1_matrix <- y[j1, , idx_3rd_dim]
      # Here, the result of "solve(...)" is a (p times dim(y)[2])-matrix or

```



```

    # a vector of length dim(y)[2] (if p == 1):
    solve(X[j2,] - X[j1,], y_j2_matrix - y_j1_matrix)
  }, simplify = "array")
  if(dim(y)[2] == 1){
    # Correction needed if dim(y)[2] == 1, so "y_j2_matrix" and
    # "y_j1_matrix" have been dropped to matrices (or even vectors,
    if also
    # p == 1):
    ee_per_3rd_dim <- array(ee_per_3rd_dim,
                          dim = c(p, dim(y)[2], dim(y)[3]))
  } else if(p == 1){
    # Correction needed if p == 1 (and dim(y)[2] > 1), so "y_j2_matr
ix" and
    # "y_j1_matrix" have been dropped to matrices:
    ee_per_3rd_dim <- array(ee_per_3rd_dim,
                          dim = c(1, dim(y)[2], dim(y)[3]))
  }
  # "ee_per_3rd_dim" is now an array of dimensions
  # c(p, dim(y)[2], dim(y)[3]). Assign the corresponding names for t
he
  # third dimension:
  if(is.null(dimnames(ee_per_3rd_dim))){
    dimnames(ee_per_3rd_dim) <- dimnames(y)
  } else{
    dimnames(ee_per_3rd_dim)[[3]] <- dimnames(y)[[3]]
  }
  return(ee_per_3rd_dim)
}
ee <- sapply(1:r, one_i_array, simplify = "array")
# Special case handling if "ee" has been dropped to a vector:
# if(is(ee,"numeric")){
if (inherits(ee, "numeric")){
  ee <- array(ee, dim = c(p, dim(y)[2], dim(y)[3], r))
  dimnames(ee) <- list(NULL, dimnames(y)[[2]], dimnames(y)[[3]], NUL
L)
}
# "ee" is an array of dimensions c(p, dim(y)[2], dim(y)[3], r), so i
t is
# transposed to an array of dimensions c(r, p, dim(y)[2], dim(y)
[3]):
ee <- aperm(ee, perm = c(4, 1, 2, 3))
}
return(ee)
}

```

## Elementary effects.

```

In [8]: z.ee <- ee.oat(z.design, z.outputs.clean)
z.ee %>% dim

```

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mu, mu\*, and sigma.

```
In [9]: z.mu <- apply(z.ee, 3, function(M){
  apply(M, 2, mean)
})
z.mu <- melt(
  cbind(
    data.table(Input=rownames(z.mu)),
    data.table(z.mu)
  ),
  id.vars="Input",
  variable.name = "Output",
  value.name = "mu"
)
z.mu %>% head
```

A data.table: 6 x 3

	Input	Output	mu
	<chr>	<fct>	<dbl>
	advertising budget	bioproduct market share mass	0.303273857
	advertising start time	bioproduct market share mass	-0.387986934
	aversion to NPV deviation	bioproduct market share mass	-0.001418284
	base external investor ask rate	bioproduct market share mass	-0.023685767
	bioproduct long term price	bioproduct market share mass	-0.160606092
	bioproduct offtake agreement	bioproduct market share mass	0.512129831

```
In [10]: z.mu.star <- apply(abs(z.ee), 3, function(M){
  apply(M, 2, mean)
})
z.mu.star <- melt(
  cbind(
    data.table(Input=rownames(z.mu.star)),
    data.table(z.mu.star)
  ),
  id.vars="Input",
  variable.name = "Output",
  value.name = "mu*"
)
z.mu.star %>% head
```

A data.table: 6 x 3

	Input	Output	mu*
	<chr>	<fct>	<dbl>
	advertising budget	bioproduct market share mass	0.31027692
	advertising start time	bioproduct market share mass	0.41195148
	aversion to NPV deviation	bioproduct market share mass	0.01152693
	base external investor ask rate	bioproduct market share mass	0.26447606
	bioproduct long term price	bioproduct market share mass	0.39980623
	bioproduct offtake agreement	bioproduct market share mass	0.77572582

```
In [11]: z.sigma <- apply(z.ee, 3, function(M){
  apply(M, 2, sd)
})
z.sigma <- melt(
  cbind(
    data.table(Input=rownames(z.sigma)),
    data.table(z.sigma)
  ),
  id.vars="Input",
  variable.name = "Output",
  value.name = "sigma"
)
z.sigma %>% head
```

A data.table: 6 x 3

	Input	Output	sigma
	<chr>	<fct>	<dbl>
	advertising budget	bioproduct market share mass	1.1085135
	advertising start time	bioproduct market share mass	1.4664525
	aversion to NPV deviation	bioproduct market share mass	0.1115146
	base external investor ask rate	bioproduct market share mass	0.6278341
	bioproduct long term price	bioproduct market share mass	1.7387415
	bioproduct offtake agreement	bioproduct market share mass	4.5942430

```
In [12]: z.results <- merge(merge(z.mu, z.mu.star, on=c("Input", "Output")), z.si
gma, on=c("Input", "Output"))
z.results %>% head
```

A data.table: 6 x 5

	Input	Output	mu	mu*	sigma
	<chr>	<fct>	<dbl>	<dbl>	<dbl>
	advertising budget	bioproduct market share mass	3.032739e-01	3.102769e-01	1.108513e+00
	advertising budget	current market size economic	-2.412707e+11	9.415770e+11	4.866741e+12
	advertising budget	current market size mass	0.000000e+00	0.000000e+00	0.000000e+00
	advertising budget	long term market share	-6.742848e-04	1.255821e-03	1.507683e-02
	advertising budget	long term market value	-3.252294e+08	3.252294e+08	7.252179e+09
	advertising budget	Adopters	4.651019e+07	5.092098e+07	1.587423e+08

```
In [13]: z.results[, `:=`(
  `mu rank` = frank(- `mu` ),
  `mu* rank` = frank(- `mu*` ),
  `sigma rank` = frank(- `sigma` )
), by=.(Output)]
z.results %>% head
```

A data.table: 6 x 8

Input	Output	mu	mu*	sigma	mu rank	mu* rank	sigma rank
<chr>	<fct>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
advertising budget	bioproduct market share mass	3.032739e-01	3.102769e-01	1.108513e+00	3	15	24
advertising budget	current market size economic	-2.412707e+11	9.415770e+11	4.866741e+12	80	23	25
advertising budget	current market size mass	0.000000e+00	0.000000e+00	0.000000e+00	43	44	44
advertising budget	long term market share	-6.742848e-04	1.255821e-03	1.507683e-02	64	49	51
advertising budget	long term market value	-3.252294e+08	3.252294e+08	7.252179e+09	71	30	24
advertising budget	Adopters	4.651019e+07	5.092098e+07	1.587423e+08	6	13	23

## Interpret results.

Interpretations:

- mu: influence of variable
- mustar: influence of variable, accounting for non-monotonicity
- sigma: non-linear and interaction effects for variable

Heat map for mu\*.

```

In [14]: ggplot(
          z.results,
          aes(
            x=factor(
              Output,
              levels=z.results[, .(`sort` = mean(`mu* rank`)), by=.(Output
)] [order(sort), `Output`]
            ),
            y=factor(
              Input,
              levels=z.results[, .(`sort` = mean(`mu* rank`)), by=.(Input)
)] [order(sort), `Input`]
            ),
            fill=`mu* rank`
          )
        ) +
        geom_tile() +
        scale_fill_distiller(type="div") +
        xlab("Input variable sorted by mean rank of mu*") +
        ylab("Output variable sorted by mean rank of mu*") +
        theme(
          axis.text=element_text(size=5),
          axis.text.x = element_text(angle = 45, hjust=1)
        )

```

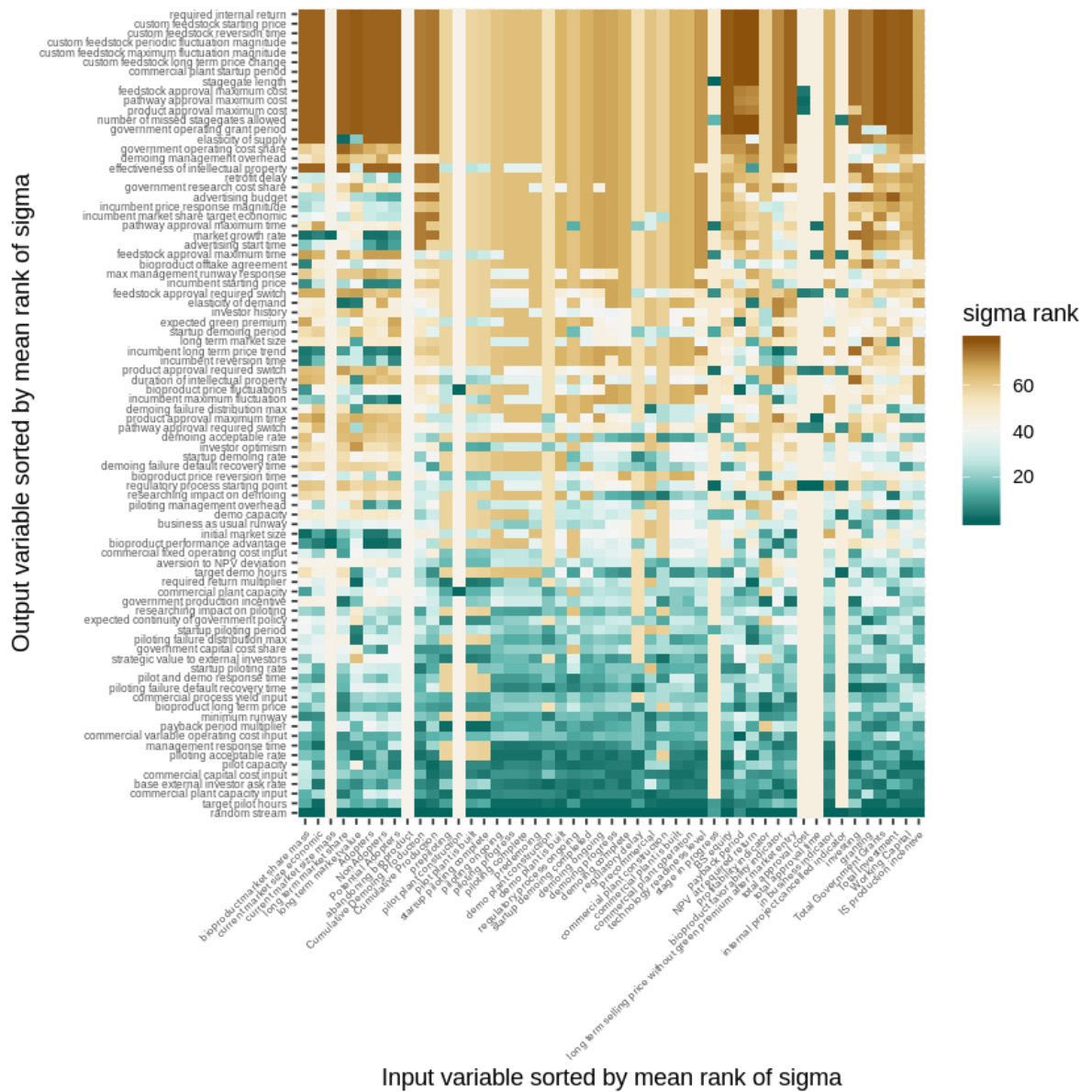


```

In [15]: ggplot(
          z.results,
          aes(
            x=factor(
              Output,
              levels=z.results[, .(`sort` = mean(`sigma rank`)), by=.(Outp
ut)][order(sort), `Output`]
            ),
            y=factor(
              Input,
              levels=z.results[, .(`sort` = mean(`sigma rank`)), by=.(Inpu
t)][order(sort), `Input`]
            ),
            fill=`sigma rank`
          )
        ) +
  geom_tile() +
  scale_fill_distiller(type="div") +
  xlab("Input variable sorted by mean rank of sigma") +
  ylab("Output variable sorted by mean rank of sigma") +
  theme(
    axis.text=element_text(size=5),
    axis.text.x = element_text(angle = 45, hjust=1)
  )

```



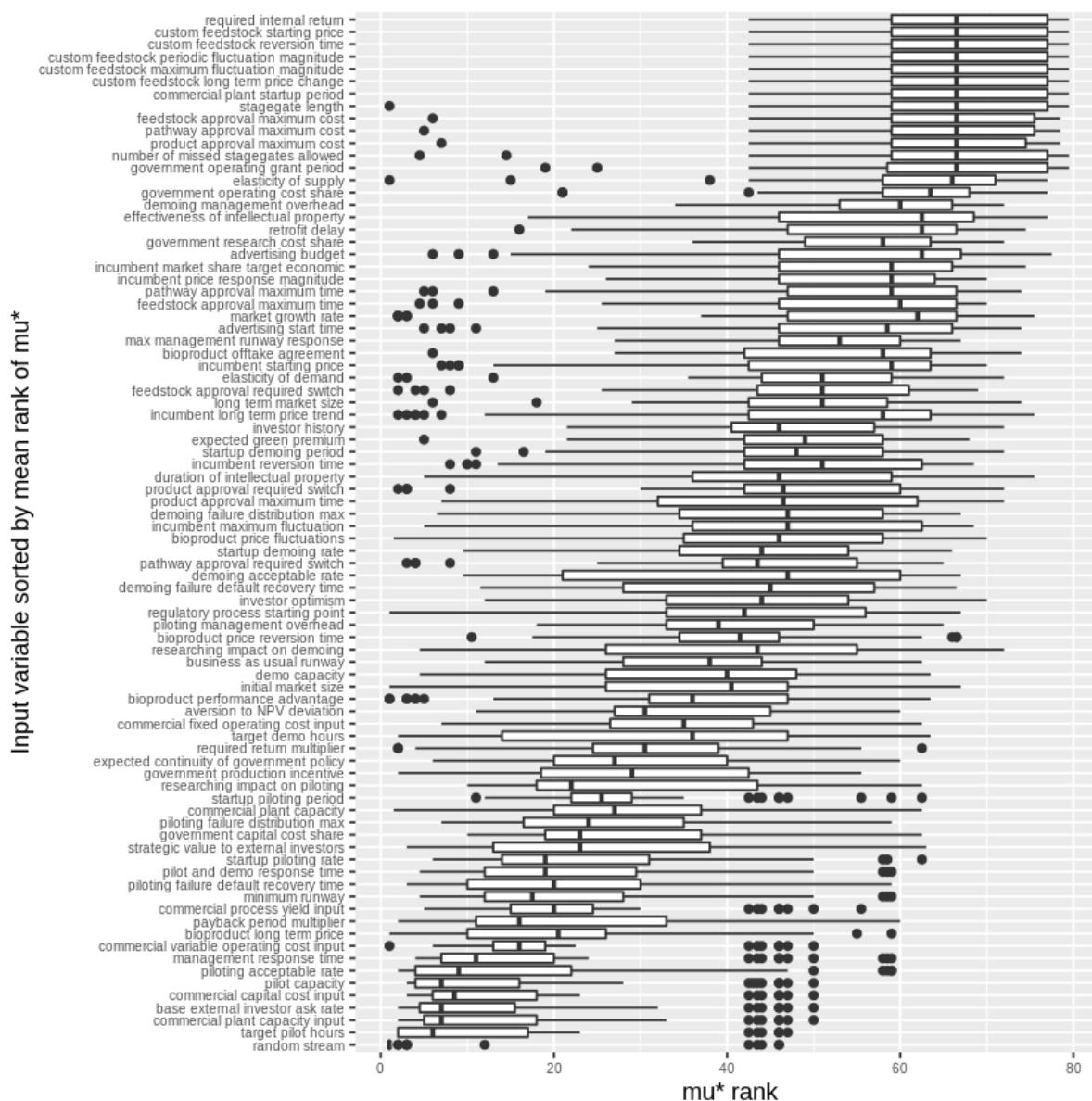


Box plots of ranks for  $\mu^*$ .

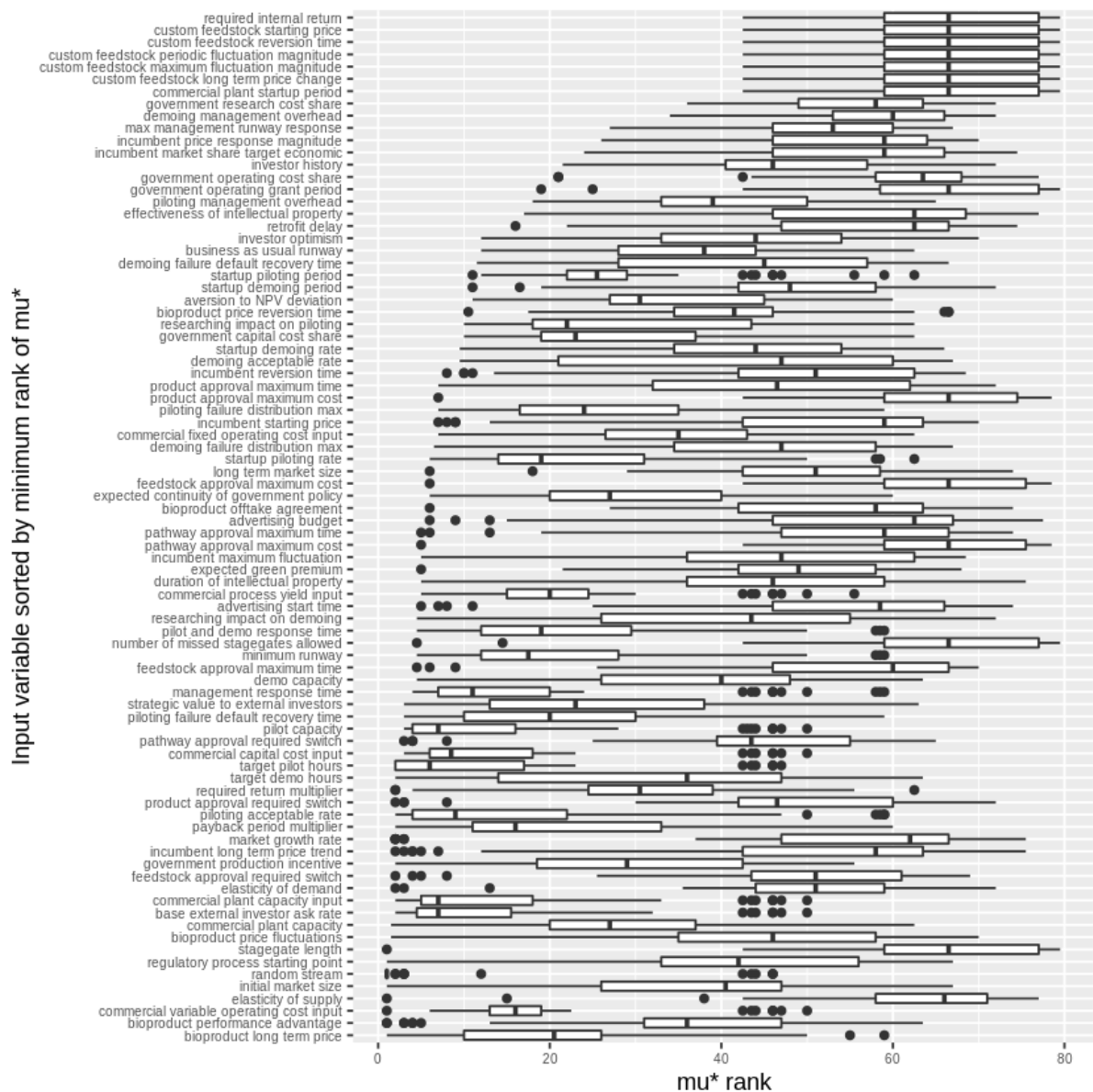
```

In [16]: ggplot(
  z.results,
  aes(
    x=factor(
      Input,
      levels=z.results[, .(`sort` = mean(`mu* rank`)), by=.(Input)
    ][order(sort), `Input`]
  ),
  y=`mu* rank`
) +
  geom_boxplot() +
  coord_flip() +
  xlab("Input variable sorted by mean rank of mu*") +
  theme(axis.text=element_text(size=6))

```



```
In [17]: ggplot(
  z.results,
  aes(
    x=factor(
      Input,
      levels=z.results[, .(`sort` = min(`mu* rank`)), by=.(Input)]
[order(sort), `Input`]
    ),
    y=`mu* rank`
  )
) +
  geom_boxplot() +
  coord_flip() +
  xlab("Input variable sorted by minimum rank of mu*") +
  theme(axis.text=element_text(size=6))
```

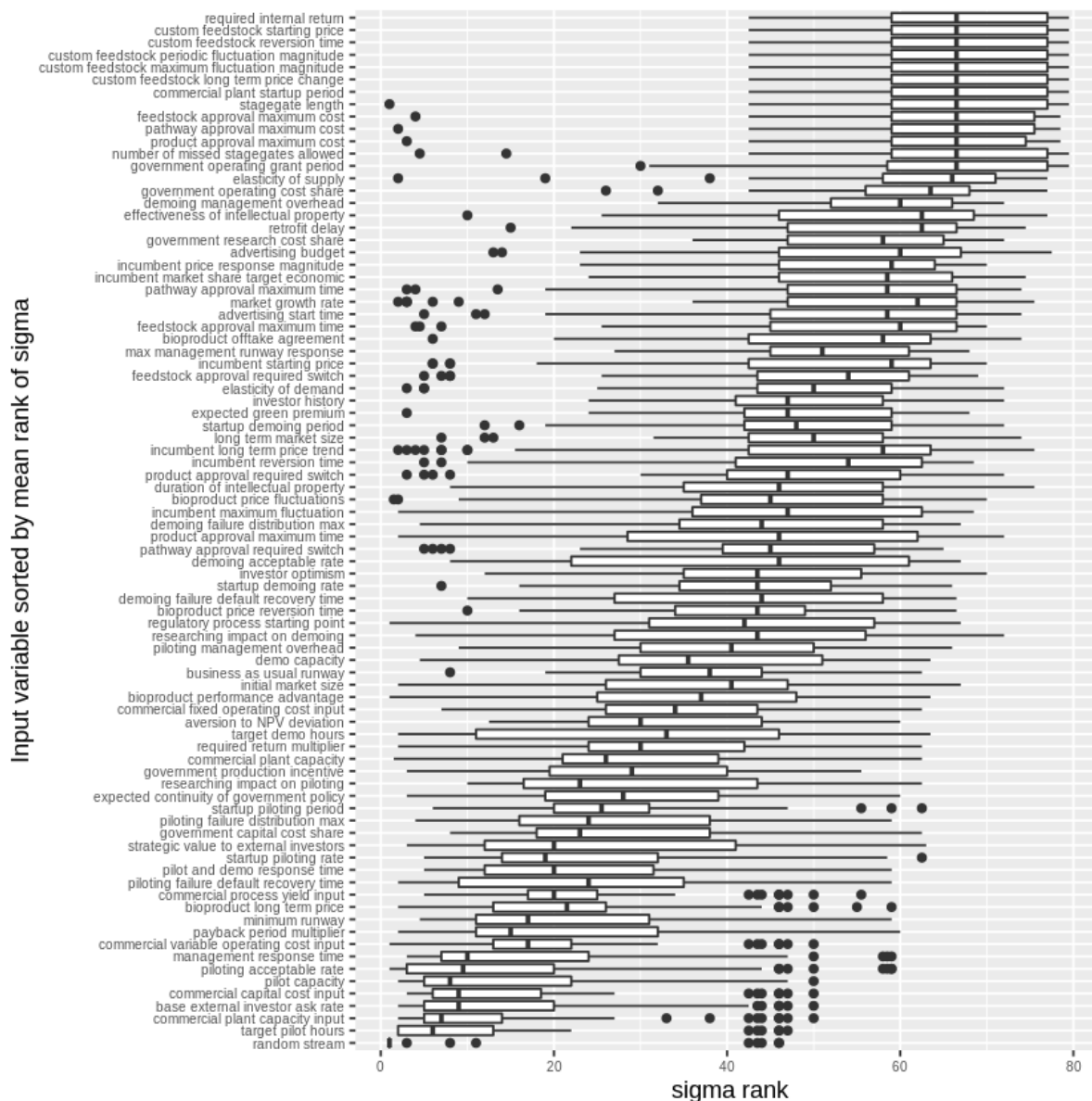


Box plots of ranks for sigma.

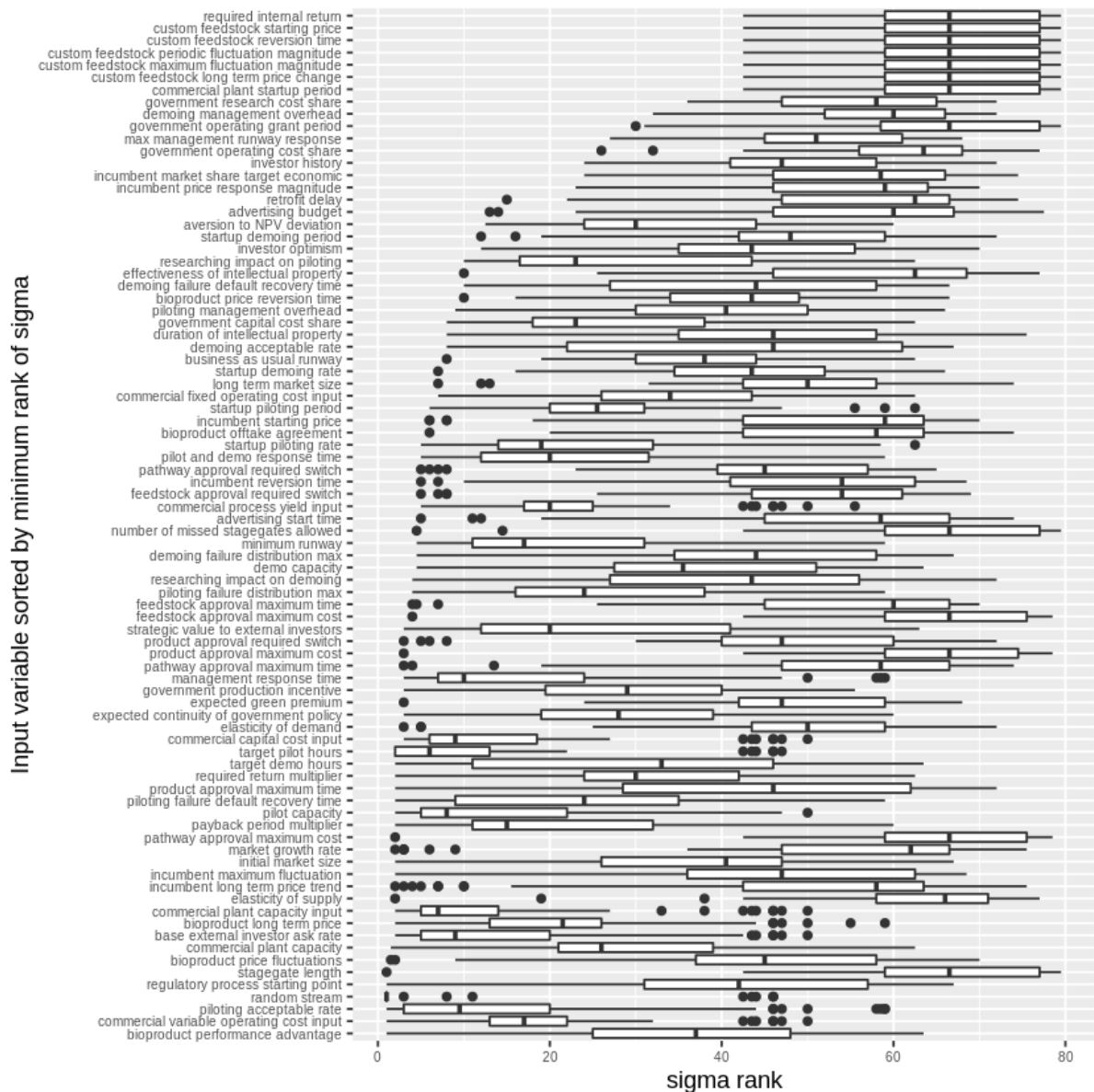
```

In [18]: ggplot(
  z.results,
  aes(
    x=factor(
      Input,
      levels=z.results[, .(`sort` = mean(`sigma rank`)), by=.(Input
t)][order(sort), `Input`]
    ),
    y=`sigma rank`
  )
) +
  geom_boxplot() +
  coord_flip() +
  xlab("Input variable sorted by mean rank of sigma") +
  theme(axis.text=element_text(size=6))

```



```
In [19]: ggplot(
  z.results,
  aes(
    x=factor(
      Input,
      levels=z.results[, .(`sort` = min(`sigma rank`)), by=.(Input
    )][order(sort), `Input`]
  ),
  y=`sigma rank`
) +
  geom_boxplot() +
  coord_flip() +
  xlab("Input variable sorted by minimum rank of sigma") +
  theme(axis.text=element_text(size=6))
```



**Select inputs variables for variance-based sensitivity analysis.**

**Select the variables whose median rank is less than 30.**

```

In [20]: z.results[,
      .(
        `mu* minimum rank` = min  (`mu* rank`),
        `mu* mean rank`    = mean  (`mu* rank`),
        `mu* median rank`  = median(`mu* rank`),
        `sigma minimum rank` = min  (`sigma rank`),
        `sigma mean rank`   = mean  (`sigma rank`),
        `sigma median rank` = median(`sigma rank`)
      ),
      by=.(Input)
][order(`mu* minimum rank`)][
  `mu* median rank` <= 30 | `sigma median rank` <= 30
]

```

A data.table: 26 x 7

Input	mu* minimum rank	mu* mean rank	mu* median rank	sigma minimum rank	sigma mean rank	sigma median rank
<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
bioproduct long term price	1.0	21.989796	20.5	2.0	22.938776	21.5
commercial variable operating cost input	1.0	19.183673	16.0	1.0	20.000000	17.0
random stream	1.0	5.816327	1.0	1.0	5.816327	1.0
commercial plant capacity	1.5	28.193878	27.0	1.5	29.418367	26.0
base external investor ask rate	2.0	13.744898	7.0	2.0	14.755102	9.0
commercial plant capacity input	2.0	13.673469	7.0	2.0	14.428571	7.0
government production incentive	2.0	28.897959	29.0	3.0	29.295918	29.0
payback period multiplier	2.0	22.020408	16.0	2.0	21.285714	15.0
piloting acceptable rate	2.0	17.908163	9.0	1.0	17.826531	9.5
required return multiplier	2.0	30.326531	30.5	2.0	30.285714	30.0
target pilot hours	2.0	11.948980	6.0	2.0	11.775510	6.0
commercial capital cost input	3.0	14.326531	8.5	3.0	15.173469	9.0
pilot capacity	3.0	14.632653	7.0	2.0	15.285714	8.0
piloting failure default recovery time	3.0	23.000000	20.0	2.0	23.857143	24.0
strategic value to external investors	3.0	26.132653	23.0	3.0	26.265306	20.0
management response time	4.0	18.091837	11.0	3.0	18.897959	10.0
minimum runway	4.5	22.683673	17.5	4.5	22.785714	17.0
pilot and demo response time	4.5	23.581633	19.0	5.0	24.030612	20.0
commercial process yield input	5.0	22.500000	20.0	5.0	22.938776	20.0
expected continuity of government policy	6.0	29.071429	27.0	3.0	28.653061	28.0
startup piloting rate	6.0	24.122449	19.0	5.0	24.377551	19.0
piloting failure distribution max	7.0	27.571429	24.0	4.0	27.397959	24.0
government capital cost share	10.0	27.520408	23.0	8.0	27.357143	23.0
researching impact on piloting	10.0	28.571429	22.0	10.0	28.755102	23.0



Input	mu* minimum rank	mu* mean rank	mu* median rank	sigma minimum rank	sigma mean rank	sigma median rank
<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
aversion to NPV deviation	11.0	34.142857	30.5	12.5	33.285714	30.0
startup piloting period	11.0	28.459184	25.5	6.0	28.020408	25.5