## Biased Experiment

```
library(conflicted)
library(kableExtra)
library(knitr)
library(broom.helpers)
library(broom)
library(dtplyr)
library(furrr)
## Loading required package: future
library(arrow)
library(glue)
library(fs)
library(tidyverse)
## -- Attaching core tidyverse packages ----
                                                   ----- tidyverse 2.0.0 --
## v dplyr
           1.1.4 v readr
                                    2.1.5
## v forcats 1.0.0 v stringr 1.5.1
## v ggplot2 3.5.1
                                    3.2.1
                        v tibble
## v lubridate 1.9.3
                        v tidyr
                                    1.3.1
## v purrr
              1.0.2
conflict_prefer("filter", "dplyr")
## [conflicted] Will prefer dplyr::filter over any other package.
source(here("analysis/utils.R"), local = knit_global())
set theme()
write_bib(.packages(), here("analysis/packages.bib"))
sessionInfo()
## R version 4.4.0 (2024-04-24)
## Platform: aarch64-apple-darwin20
## Running under: macOS Sonoma 14.5
## Matrix products: default
         /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib; LAPACK v
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
## time zone: Asia/Singapore
## tzcode source: internal
## attached base packages:
## [1] stats
              graphics grDevices utils
                                              datasets methods
                                                                  base
```

```
##
## other attached packages:
## [1] lubridate_1.9.3
                             forcats 1.0.0
                                                   stringr 1.5.1
## [4] dplyr_1.1.4
                             purrr_1.0.2
                                                   readr_2.1.5
## [7] tidyr_1.3.1
                             tibble_3.2.1
                                                   ggplot2_3.5.1
## [10] tidyverse 2.0.0
                             fs 1.6.4
                                                   glue_1.7.0
## [13] arrow_16.1.0
                             furrr_0.3.1
                                                   future 1.33.2
## [16] dtplyr_1.3.1
                             broom_1.0.6
                                                   broom.helpers_1.15.0
## [19] knitr_1.47
                             kableExtra_1.4.0
                                                   conflicted_1.2.0
## [22] here_1.0.1
## loaded via a namespace (and not attached):
## [1] gtable_0.3.5
                          xfun_0.45
                                             tzdb_0.4.0
                                                               vctrs_0.6.5
## [5] tools_4.4.0
                          generics_0.1.3
                                             parallel_4.4.0
                                                               fansi_1.0.6
                          data.table_1.15.4 assertthat_0.2.1
## [9] pkgconfig_2.0.3
                                                               lifecycle_1.0.4
## [13] compiler_4.4.0
                          munsell_0.5.1
                                             codetools_0.2-20
                                                               htmltools_0.5.8.1
## [17] yaml_2.3.8
                          pillar_1.9.0
                                             cachem_1.1.0
                                                               parallelly_1.37.1
## [21] tidyselect_1.2.1 digest_0.6.35
                                             stringi_1.8.4
                                                               listenv 0.9.1
## [25] rprojroot_2.0.4
                          fastmap_1.2.0
                                             grid_4.4.0
                                                               colorspace_2.1-0
## [29] cli_3.6.2
                          magrittr_2.0.3
                                            utf8_1.2.4
                                                               withr 3.0.0
## [33] scales_1.3.0
                          backports_1.5.0
                                            bit64_4.0.5
                                                               timechange_0.3.0
## [37] rmarkdown 2.27
                          globals_0.16.3
                                             bit_4.0.5
                                                               hms_1.1.3
## [41] memoise_2.0.1
                          evaluate_0.24.0
                                             viridisLite_0.4.2 rlang_1.1.4
## [45] xml2 1.3.6
                          svglite_2.1.3
                                             rstudioapi_0.16.0 R6_2.5.1
## [49] systemfonts_1.1.0
```

## Analyze attack trends

```
data_dir <- here(glue("{params$data}/{params$simulation}/results"))
success_fnames <-
    dir_ls(data_dir, glob = glue("*norm_{params$norm}*.csv"))
stopifnot(length(success_fnames) == 240)
# every fname is a simulation
success_raw_data <- get_data(success_fnames, read_csv) |>
    glimpse()
```

```
## Rows: 240
## Columns: 16
## $ fname
                         <chr> "/Users/zbli/Documents/Documents - ZhaoBin's M~
                         ## $ num iteration
## $ max_norm
                         <dbl> 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05~
## $ model name
                         <ord> Cascade R-CNN, Faster R-CNN, RetinaNet, SSD, Y~
                         <ord> Mislabeling, Mislabeling, Mislabe-
## $ loss_target
                         <chr> "predictions", "predictions", "predictions", "~
## $ attack_bbox
                         <chr> "perturb_inside", "perturb_inside", "perturb_i~
## $ perturb_fun
## $ sample_count
                         <dbl> 1258, 1301, 703, 1105, 1157, 1258, 1301, 703, ~
                         ## $ attack_count
## $ success_count
                         <dbl> 25, 23, 13, 62, 86, 47, 61, 58, 52, 51, 31, 25~
                         <dbl> 15, 14, 5, 8, 39, 44, 59, 58, 48, 46, 30, 24, ~
## $ vanish_count
## $ mislabel_count
                         <dbl> 10, 9, 8, 54, 47, 3, 2, 0, 4, 5, 1, 1, 1, 1, 2~
```

```
## $ mislabel_intended_count <dbl> 10, 9, 8, 54, 47, 0, 0, 0, 0, 0, 0, 0, 0, 0 ~ 0~
                       ## $ target_max_conf
                       <dbl> 0.25, 0.25, 0.25, 0.25, 0.25, 0.25, 0.25, 0.25
## $ perturb min size
                       <dbl> 0.25, 0.25, 0.25, 0.25, 0.25, 0.25, 0.25, 0.25
## $ bbox_max_dist
# target_max_conf, perturb_min_size, bbox_max_dist are the sampling criteria
success_raw_data <- success_raw_data |>
 rowwise() |>
 mutate(across(target_max_conf:bbox_max_dist, ~ !is.na(.)), # convert to TRUE/FALSE
   num_cri = sum(across(target_max_conf:bbox_max_dist))
 ) |>
 glimpse()
## Rows: 240
## Columns: 17
## Rowwise:
## $ fname
                       <chr> "/Users/zbli/Documents/Documents - ZhaoBin's M~
                       ## $ num iteration
## $ max norm
                       <dbl> 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05~
## $ model name
                       <ord> Cascade R-CNN, Faster R-CNN, RetinaNet, SSD, Y~
                       <ord> Mislabeling, Mislabeling, Mislabe-
## $ loss_target
                       <chr> "predictions", "predictions", "predictions", "~
## $ attack bbox
                       <chr> "perturb_inside", "perturb_inside", "perturb_i~
## $ perturb fun
## $ sample_count
                       <dbl> 1258, 1301, 703, 1105, 1157, 1258, 1301, 703, ~
## $ attack count
                       <dbl> 25, 23, 13, 62, 86, 47, 61, 58, 52, 51, 31, 25~
## $ success_count
                       <dbl> 15, 14, 5, 8, 39, 44, 59, 58, 48, 46, 30, 24, ~
## $ vanish_count
                       <dbl> 10, 9, 8, 54, 47, 3, 2, 0, 4, 5, 1, 1, 1, 1, 2~
## $ mislabel_count
## $ mislabel_intended_count <dbl> 10, 9, 8, 54, 47, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0
                       <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
## $ target_max_conf
                       <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
## $ perturb_min_size
                       <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
## $ bbox_max_dist
## $ num_cri
                       # expand success per simulation into 1 and 0s per row
success_expanded_data <- success_raw_data |>
 rowwise() |>
 mutate(success = list(rep(0:1, times = c(attack_count - success_count, success_count)))) |>
 unnest longer(success) |>
 glimpse()
## Rows: 24,000
## Columns: 18
                       <chr> "/Users/zbli/Documents/Documents - ZhaoBin's M~
## $ fname
                       ## $ num_iteration
## $ max_norm
                       <dbl> 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05~
## $ model_name
                       <ord> Cascade R-CNN, Cascade R-CNN, Cascade R-CNN, C~
                       <ord> Mislabeling, Mislabeling, Mislabe
## $ loss_target
## $ attack_bbox
                       <chr> "predictions", "predictions", "predictions", "~
                       <chr> "perturb_inside", "perturb_inside", "perturb_i~
## $ perturb_fun
## $ sample count
                       <dbl> 1258, 1258, 1258, 1258, 1258, 1258, 1258, 1258,
                       ## $ attack_count
## $ success count
                       ## $ vanish_count
                       ## $ mislabel_count
```

```
<lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
## $ target max conf
                        <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
## $ perturb_min_size
## $ bbox_max_dist
                        <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
                        ## $ num_cri
## $ success
                        itr lab <- "Number of Factors"</pre>
cap <- glue("{emp_tex('Success factors can be exploited in combination to significantly increase succes
cap
## Success factors can be exploited in combination to significantly increase success rates even with 0.
# use linear
g <- success_expanded_data |>
 ggplot(aes(num_cri, success, color = loss_target, linetype = loss_target)) +
 # use stat_summary rather than stat_summary_bin
 # since num_cri is set experimentally
 # mean cl boot gives 95% bootstrapped CI at 1000 samples
 # https://rdrr.io/cran/Hmisc/man/smean.sd.html
 stat_summary(fun.data = "mean_cl_boot") +
 binomial_smooth(formula = y ~ x) +
 facet_grid(cols = vars(model_name))
g +
 labs(x = itr_lab, y = glue("p(Success) {norm_axy(params$norm)}"), color = "Attack", linetype = "Attack"
 scale_x_continuous(breaks = unique(success_raw_data$num_cri))
data <- success_expanded_data |>
 # avoid ordered regression
 mutate(
   model_name = factor(model_name, ordered = FALSE),
   loss_target = factor(loss_target, ordered = FALSE)
 ) |>
 glimpse()
## Rows: 24,000
## Columns: 18
## $ fname
                        <chr> "/Users/zbli/Documents/Documents - ZhaoBin's M~
                        ## $ num iteration
## $ max norm
                        <dbl> 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05, 0.05~
                        <fct> Cascade R-CNN, Cascade R-CNN, Cascade R-CNN, C~
## $ model_name
## $ loss_target
                        <fct> Mislabeling, Mislabeling, Mislabeling, Mislabe~
                        <chr> "predictions", "predictions", "predictions", "~
## $ attack_bbox
                        <chr> "perturb_inside", "perturb_inside", "perturb_i~
## $ perturb_fun
                        <dbl> 1258, 1258, 1258, 1258, 1258, 1258, 1258, 1258,
## $ sample_count
## $ attack_count
                        ## $ success_count
## $ vanish_count
                        ## $ mislabel_count
## $ target max conf
                        <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
## $ perturb_min_size
                        <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
                        <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
## $ bbox max dist
```

## \$ num\_cri

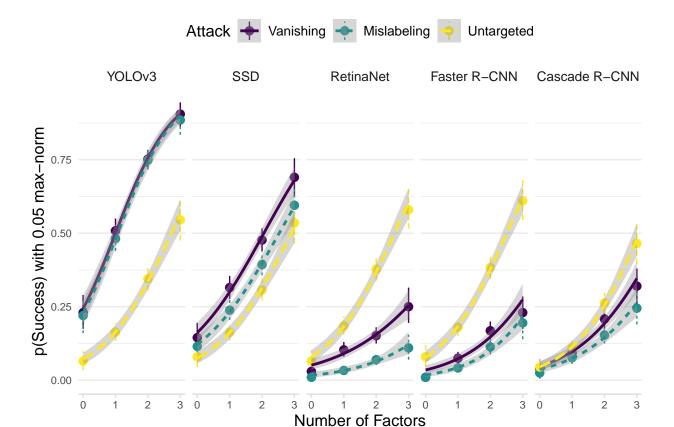


Figure 1: Success factors can be exploited in combination to significantly increase success rates even with 0.05 max-norm: We sampled target and perturb objects based on three validated success factors in Table ?? by targeting objects with low predicted confidence, perturbing large objects and selecting target and perturb objects close to one another. The binned summaries and regression trendlines graph success proportion against number of factors in the deliberate attack experiment. Errors are 95% confidence intervals and every point aggregates success over 200 images. Success rates significantly increase as the number of factors combined increases. Significance is determined at  $\alpha < 0.05$  using a Wald z-test on the logistic estimates. Full details are given in Section ??.

```
## $ success
model <- partial(glm model, predictor = "num cri")</pre>
reg_est <- get_tidied_reg(</pre>
 model, data
)
## `summarise()` has grouped output by 'model_name', 'loss_target'. You can
## override using the `.groups` argument.
ext_sig(reg_est, "pos")
## Total 15 predictors:
## 15 (100%) significant;
## 15 (100%) pos
## # A tibble: 15 x 9
## # Groups:
             model_name, loss_target [15]
##
     model name
                  loss_target term estimate std.error statistic p.value conf.low
```

```
##
      <fct>
                     <fct>
                                  <chr>
                                            <dbl>
                                                       <dbl>
                                                                 <dbl>
                                                                          <dbl>
                                                                                    <dbl>
##
    1 YOLOv3
                     Vanishing
                                            1.13
                                                       0.075
                                                                 15.1
                                                                              0
                                                                                    0.987
                                  num_~
##
    2 YOLOv3
                     Mislabeling num_~
                                            1.13
                                                       0.074
                                                                 15.2
                                                                              0
                                                                                    0.991
    3 YOLOv3
                     Untargeted num_~
                                                                              0
##
                                            0.938
                                                       0.076
                                                                 12.3
                                                                                    0.791
##
    4 SSD
                     Vanishing
                                  num ~
                                            0.8
                                                       0.067
                                                                 12.0
                                                                              0
                                                                                    0.671
##
    5 SSD
                     Mislabeling num ~
                                                       0.069
                                                                 11.3
                                                                              0
                                                                                    0.647
                                            0.781
##
    6 SSD
                     Untargeted num_~
                                            0.864
                                                       0.076
                                                                 11.4
                                                                              0
                                                                                    0.718
                     Vanishing
                                  num_~
                                                                  6.76
                                                                              0
                                                                                    0.439
##
    7 RetinaNet
                                            0.615
                                                       0.091
                     Mislabeling num_~
##
    8 RetinaNet
                                            0.703
                                                       0.137
                                                                  5.13
                                                                              0
                                                                                    0.438
                                                                              0
                                                                                    0.812
##
    9 RetinaNet
                     Untargeted
                                  num_~
                                            0.956
                                                       0.075
                                                                 12.8
## 10 Faster R-CNN
                     Vanishing
                                  num_~
                                            0.783
                                                       0.097
                                                                  8.06
                                                                              0
                                                                                    0.595
                                                                  7.73
                                                                              0
## 11 Faster R-CNN
                     Mislabeling num_~
                                            0.903
                                                       0.117
                                                                                    0.678
## 12 Faster R-CNN
                     Untargeted num_~
                                            0.985
                                                       0.075
                                                                 13.1
                                                                              0
                                                                                    0.84
## 13 Cascade R-CNN Vanishing
                                            0.794
                                                       0.088
                                                                              0
                                                                                    0.625
                                                                  9.07
## 14 Cascade R-CNN Mislabeling num_~
                                            0.743
                                                       0.097
                                                                  7.69
                                                                              0
                                                                                    0.556
## 15 Cascade R-CNN Untargeted num_~
                                            0.982
                                                       0.084
                                                                 11.7
                                                                              0
                                                                                    0.82
## # i 1 more variable: conf.high <dbl>
```

cap <- table\_caption(glue("log({itr\_lab})"), "Success rates increase with the number of factors combine
print\_statistics(reg\_est, cap)</pre>

Table 1: We run a logistic model regressing success against log(number of factors) in the randomized attack experiment. Success rates increase with the number of factors combined to select target and perturb objects for all models and attacks. Table headers are explained in Appendix ??.

	Group	Regression							
	Attack	term	sig	estimate	std.error	statistic	p.value	conf.low	conf.high
YOLOv3									
	Vanishing	$num\_cri$	*	1.132	0.075	15.133	0	0.987	1.281
	Mislabeling	num_cri	*	1.134	0.074	15.240	0	0.991	1.283
	Untargeted	num_cri	*	0.938	0.076	12.307	0	0.791	1.090
SSI	D								
	Vanishing	$num\_cri$	*	0.800	0.067	11.974	0	0.671	0.933
	Mislabeling	num_cri	*	0.781	0.069	11.291	0	0.647	0.919
	Untargeted	num_cri	*	0.864	0.076	11.408	0	0.718	1.015
Ret	tinaNet								
	Vanishing	$num\_cri$	*	0.615	0.091	6.760	0	0.439	0.795
	Mislabeling	num_cri	*	0.703	0.137	5.131	0	0.438	0.976
	Untargeted	num_cri	*	0.956	0.075	12.780	0	0.812	1.105
Fas	ster R-CNN								
	Vanishing	$num\_cri$	*	0.783	0.097	8.062	0	0.595	0.976
	Mislabeling	num_cri	*	0.903	0.117	7.729	0	0.678	1.136
	Untargeted	num_cri	*	0.985	0.075	13.115	0	0.840	1.135
Cascade R-CNN									
	Vanishing	num_cri	*	0.794	0.088	9.069	0	0.625	0.968
	Mislabeling	num_cri	*	0.743	0.097	7.689	0	0.556	0.936
	Untargeted	num_cri	*	0.982	0.084	11.703	0	0.820	1.149

```
success_expanded_data |>
 group_by(model_name, loss_target, num_cri) |>
 summarize(mean(success))
## `summarise()` has grouped output by 'model_name', 'loss_target'. You can
## override using the `.groups` argument.
## # A tibble: 60 x 4
## # Groups: model_name, loss_target [15]
     model_name loss_target num_cri `mean(success)`
##
     <ord>
              <ord>
                            <int>
                                           <dbl>
## 1 YOLOv3
               Vanishing
                            0
                                           0.23
## 2 YOLOv3 Vanishing
                                          0.508
                               1
## 3 YOLOv3
               Vanishing
                              2
                                           0.752
            Vanishing
Mislabeling
                                           0.905
## 4 YOLOv3
                               3
## 5 YOLOv3
                               0
                                           0.22
## 6 YOLOv3
            Mislabeling
                              1
                                           0.482
## 7 YOLOv3
            Mislabeling
                              2
                                           0.748
## 8 YOLOv3
                              3
                                           0.885
               Mislabeling
## 9 YOLOv3
               Untargeted
                               0
                                           0.065
                              1
## 10 YOLOv3
               Untargeted
                                           0.163
## # i 50 more rows
```