

Deliberate Attack

```
library(conflicted)

library(kableExtra)

## Warning in !is.null(rmarkdown::metadata$output) && rmarkdown::metadata$output
## %in% : 'length(x) = 2 > 1' in coercion to 'logical(1)'

library(knitr)
library(broom.helpers)
library(broom)
library(dtplyr)
library(furrr)

## Loading required package: future

library(arrow)
library(glue)
library(fs)
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.2 --

## v ggplot2 3.4.0      v purrr   1.0.1
## v tibble  3.1.8      v dplyr  1.0.10
## v tidyr   1.2.1      v stringr 1.5.0
## v readr   2.1.3      v forcats 0.5.2

conflict_prefer("filter", "dplyr")

## [conflicted] Will prefer dplyr::filter over any other package

source("./analysis/utils.R", local = knitr_global())
set_theme()

write_bib(.packages(), glue("./analysis/packages.bib"))

## Warning in utils::citation(..., lib.loc = lib.loc): no date field in DESCRIPTION
## file of package 'kableExtra'

sessionInfo()

## R version 4.2.2 (2022-10-31)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS Big Sur ... 10.16
##
## Matrix products: default
## BLAS:   /Library/Frameworks/R.framework/Versions/4.2/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.2/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
```

```
##
## attached base packages:
## [1] stats      graphics  grDevices datasets  utils      methods    base
##
## other attached packages:
## [1] forcats_0.5.2      stringr_1.5.0      dplyr_1.0.10
## [4] purrr_1.0.1        readr_2.1.3        tidyr_1.2.1
## [7] tibble_3.1.8       ggplot2_3.4.0      tidyverse_1.3.2
## [10] fs_1.5.2           glue_1.6.2         arrow_10.0.1
## [13] furrr_0.3.1        future_1.30.0      dtplyr_1.2.2
## [16] broom_1.0.2         broom.helpers_1.11.0 knitr_1.41
## [19] kableExtra_1.3.4.9000 conflicted_1.1.0
##
## loaded via a namespace (and not attached):
## [1] httr_1.4.4          bit64_4.0.5         jsonlite_1.8.4
## [4] viridisLite_0.4.1  here_1.0.1          modelr_0.1.10
## [7] assertthat_0.2.1   renv_0.16.0         googlesheets4_1.0.1
## [10] cellranger_1.1.0   yaml_2.3.6          globals_0.16.2
## [13] pillar_1.8.1       backports_1.4.1     digest_0.6.31
## [16] rvest_1.0.3        colorspace_2.0-3    htmltools_0.5.4
## [19] pkgconfig_2.0.3    listenv_0.9.0       haven_2.5.1
## [22] scales_1.2.1       webshot_0.5.4       svglite_2.1.1
## [25] tzdb_0.3.0         timechange_0.2.0    googledrive_2.0.0
## [28] generics_0.1.3     ellipsis_0.3.2      withr_2.5.0
## [31] cachem_1.0.6       cli_3.6.0           crayon_1.5.2
## [34] readxl_1.4.1       magrittr_2.0.3      memoise_2.0.1
## [37] evaluate_0.19      fansi_1.0.3         parallelly_1.34.0
## [40] xml2_1.3.3         tools_4.2.2         data.table_1.14.6
## [43] hms_1.1.2          gargle_1.2.1        lifecycle_1.0.3
## [46] reprex_2.0.2       munsell_0.5.0       compiler_4.2.2
## [49] systemfonts_1.0.4  rlang_1.0.6         grid_4.2.2
## [52] rstudioapi_0.14    rmarkdown_2.19      gtable_0.3.1
## [55] codetools_0.2-18   DBI_1.1.3           R6_2.5.1
## [58] lubridate_1.9.0    fastmap_1.1.0       bit_4.0.5
## [61] utf8_1.2.2         rprojroot_2.0.3     stringi_1.7.12
## [64] parallel_4.2.2     vctrs_0.5.1         dbplyr_2.2.1
## [67] tidyselect_1.2.0   xfun_0.36

data_dir <- path(glue("./data/{params$simulation}/results"))

success_fnames <-
  dir_ls(data_dir, glob = glue("*{params$simulation}*_trend.csv"))

# every fname is a simulation
success_raw_data <- get_data(success_fnames, read_csv) |>
  glimpse()

## Rows: 480
## Columns: 14
## $ fname                <chr> "./data/arbitrary/results/bbox_100_dist_100_re-
## $ num_iteration         <dbl> 200, 200, 200, 200, 200, 200, 200, 200, 200, 2~
## $ attack_count          <dbl> 100, 100, 100, 100, 100, 100, 100, 100, 100, 1~
## $ success_count         <dbl> 0, 0, 0, 16, 36, 1, 2, 4, 12, 7, 0, 0, 1, 27, ~
## $ vanish_count         <dbl> 0, 0, 0, 7, 27, 1, 2, 4, 11, 6, 0, 0, 1, 26, 2~
## $ mislabel_count        <dbl> 0, 0, 0, 9, 9, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0~
```

```

## $ mislabel_intended_count <dbl> 0, 0, 0, 9, 9, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~
## $ sample_count          <dbl> 116, 116, 118, 118, 119, 116, 116, 118, 118, 1~
## $ model_name            <ord> Cascade R-CNN, Faster R-CNN, RetinaNet, SSD, Y~
## $ loss_target           <ord> Mislabeling, Mislabeling, Mislabeling, Mislabel~
## $ attack_bbox           <chr> "ground_truth", "ground_truth", "ground_truth"~
## $ perturb_fun           <chr> "perturb_inside", "perturb_inside", "perturb_i~
## $ arbitrary_bbox_length <dbl> 100, 100, 100, 100, 100, 100, 100, 100, 100, 1~
## $ boundary_distance     <dbl> 100, 100, 100, 100, 100, 100, 100, 100, 100, 1~

# expand success per simulation into 1 and 0s per row
success_expanded_data <- success_raw_data |>
  rename(
    bbox_dist = boundary_distance,
    bbox_len = arbitrary_bbox_length
  ) |>
  rowwise() |>
  mutate(success = list(rep(0:1, times = c(attack_count - success_count, success_count)))) |>
  unnest_longer(success) |>
  glimpse()

## Rows: 48,000
## Columns: 15
## $ fname                <chr> "./data/arbitrary/results/bbox_100_dist_100_re~
## $ num_iteration        <dbl> 200, 200, 200, 200, 200, 200, 200, 200, 200, 2~
## $ attack_count         <dbl> 100, 100, 100, 100, 100, 100, 100, 100, 100, 1~
## $ success_count        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~
## $ vanish_count        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~
## $ mislabel_count       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~
## $ mislabel_intended_count <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~
## $ sample_count         <dbl> 116, 116, 116, 116, 116, 116, 116, 116, 116, 1~
## $ model_name           <ord> Cascade R-CNN, Cascade R-CNN, Cascade R-CNN, C~
## $ loss_target          <ord> Mislabeling, Mislabeling, Mislabeling, Mislabel~
## $ attack_bbox          <chr> "ground_truth", "ground_truth", "ground_truth"~
## $ perturb_fun          <chr> "perturb_inside", "perturb_inside", "perturb_i~
## $ bbox_len             <dbl> 100, 100, 100, 100, 100, 100, 100, 100, 100, 1~
## $ bbox_dist            <dbl> 100, 100, 100, 100, 100, 100, 100, 100, 100, 1~
## $ success              <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~

# check whether attack count equals experiment settings
stopifnot(all(success_raw_data$attack_count == 100))

reps <- success_raw_data |>
  count(model_name, loss_target, arbitrary_bbox_length, boundary_distance) |>
  glimpse()

## Rows: 240
## Columns: 5
## $ model_name           <ord> YOLOv3, YOLOv3, YOLOv3, YOLOv3, YOLOv3, YOLOv3, ~
## $ loss_target          <ord> Vanishing, Vanishing, Vanishing, Vanishing, Vani~
## $ arbitrary_bbox_length <dbl> 10, 10, 10, 10, 50, 50, 50, 50, 100, 100, 100, 1~
## $ boundary_distance    <dbl> 10, 50, 100, 200, 10, 50, 100, 200, 10, 50, 100, ~
## $ n                   <int> 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, ~

stopifnot(unique(reps$n) == 2)

```

```

# control both
model <- partial(glm_model, predictor = "bbox_dist * bbox_len")
data <- success_expanded_data

reg_res <- get_tidied_reg(model, data, return_mod = TRUE)

## `summarise()` has grouped output by 'model_name', 'loss_target'. You can
## override using the `.groups` argument.

reg_est <- reg_res$tidied

ext_sig(reg_est, "neg", "bbox_dist")

## -----bbox_dist-----
## Total 15 predictors:
## 15 (100%) significant;
## 15 (100%) neg

## # A tibble: 15 x 9
## # Groups:   model_name, loss_target [15]
##   model_name    loss_ta~1 term  estim~2 std.e~3 stati~4 p.value conf.~5 conf.~6
##   <ord>         <ord>    <chr>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 YOLOv3        Vanishing bbox~ -0.015  0.002   -7.68      0 -0.018 -0.011
## 2 YOLOv3        Mislabel~ bbox~ -0.015  0.002   -8.54      0 -0.018 -0.011
## 3 YOLOv3        Untarget~ bbox~ -0.021  0.003   -7.44      0 -0.026 -0.015
## 4 SSD           Vanishing bbox~ -0.015  0.002   -7.06      0 -0.019 -0.011
## 5 SSD           Mislabel~ bbox~ -0.018  0.002   -7.55      0 -0.023 -0.014
## 6 SSD           Untarget~ bbox~ -0.018  0.002   -7.74      0 -0.023 -0.014
## 7 RetinaNet     Vanishing bbox~ -0.045  0.006   -7.19      0 -0.058 -0.033
## 8 RetinaNet     Mislabel~ bbox~ -0.031  0.007   -4.24      0 -0.047 -0.018
## 9 RetinaNet     Untarget~ bbox~ -0.038  0.005   -7.45      0 -0.049 -0.029
## 10 Faster R-CNN Vanishing bbox~ -0.061  0.01    -6.41      0 -0.081 -0.044
## 11 Faster R-CNN Mislabel~ bbox~ -0.054  0.012   -4.66      0 -0.08  -0.034
## 12 Faster R-CNN Untarget~ bbox~ -0.044  0.006   -8.01      0 -0.056 -0.034
## 13 Cascade R-CNN Vanishing bbox~ -0.063  0.01    -6.58      0 -0.083 -0.046
## 14 Cascade R-CNN Mislabel~ bbox~ -0.062  0.012   -5.24      0 -0.088 -0.041
## 15 Cascade R-CNN Untarget~ bbox~ -0.061  0.008   -7.54      0 -0.079 -0.047
## # ... with abbreviated variable names 1: loss_target, 2: estimate,
## #   3: std.error, 4: statistic, 5: conf.low, 6: conf.high

ext_sig(reg_est, "pos", "bbox_len")

## -----bbox_len-----
## Total 15 predictors:
## 14 (93%) significant;
## 14 (93%) pos

## # A tibble: 14 x 9
## # Groups:   model_name, loss_target [14]
##   model_name    loss_ta~1 term  estim~2 std.e~3 stati~4 p.value conf.~5 conf.~6
##   <ord>         <ord>    <chr>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 YOLOv3        Vanishing bbox~  0.03    0.002   19.6      0  0.027  0.033
## 2 YOLOv3        Mislabel~ bbox~  0.019  0.001   16.6      0  0.016  0.021
## 3 YOLOv3        Untarget~ bbox~  0.007  0.001    6.53     0  0.005  0.009
## 4 SSD           Vanishing bbox~  0.024  0.001   17.7      0  0.021  0.027
## 5 SSD           Mislabel~ bbox~  0.02    0.001   16.0      0  0.017  0.022

```

```
## 6 SSD Untarget~ bbox~ 0.013 0.001 11.7 0 0.011 0.015
## 7 RetinaNet Vanishing bbox~ 0.016 0.001 10.6 0 0.013 0.019
## 8 RetinaNet Mislabel~ bbox~ 0.008 0.002 4.54 0 0.005 0.012
## 9 RetinaNet Untarget~ bbox~ 0.005 0.001 3.97 0 0.003 0.008
## 10 Faster R-CNN Vanishing bbox~ 0.011 0.001 7.13 0 0.008 0.014
## 11 Faster R-CNN Mislabel~ bbox~ 0.007 0.002 3.71 0 0.003 0.01
## 12 Faster R-CNN Untarget~ bbox~ 0.005 0.001 3.68 0 0.002 0.007
## 13 Cascade R-CNN Vanishing bbox~ 0.015 0.002 9.40 0 0.012 0.018
## 14 Cascade R-CNN Mislabel~ bbox~ 0.01 0.002 5.80 0 0.006 0.013
## # ... with abbreviated variable names 1: loss_target, 2: estimate,
## # 3: std.error, 4: statistic, 5: conf.low, 6: conf.high
```

```
ext_sig(reg_est, "both", "bbox_dist:(bbox_len)")
```

```
## -----bbox_dist:bbox_len-----
## Total 15 predictors:
## 7 (47%) significant;
## 7 (47%) both
```

```
## # A tibble: 7 x 9
## # Groups:   model_name, loss_target [7]
##   model_name loss_tar~1 term estim~2 std.e~3 stati~4 p.value conf.~5 conf.~6
##   <ord>      <ord>      <chr>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
## 1 YOLOv3     Vanishing  bbox~      0      0   -6.08    0      0      0
## 2 SSD        Vanishing  bbox~      0      0   -4.82    0      0      0
## 3 SSD        Mislabeli~ bbox~      0      0   -2.46   0.014    0      0
## 4 RetinaNet  Vanishing  bbox~      0      0   -2.15   0.032    0      0
## 5 RetinaNet  Untargeted bbox~      0      0    6.92    0      0      0
## 6 Faster R-CNN Untargeted bbox~      0      0    6.89    0      0      0
## 7 Cascade R-CNN Untargeted bbox~      0      0    6.20    0      0      0
## # ... with abbreviated variable names 1: loss_target, 2: estimate,
## # 3: std.error, 4: statistic, 5: conf.low, 6: conf.high
```

```
dist_lab <- "Perturb-Target Distance"
len_lab <- "Perturb Box Length"
```

```
pred_name <- glue("{dist_lab} and {len_lab}")
main_pt <- glue("longer {len_lab} or shorter {dist_lab} cause success rates to significantly increase f
print_statistics(reg_est, table_caption(pred_name, main_pt, "deliberate"))
```

Table 1: We run a logistic model regressing success against perturb-target distance and perturb box length in the deliberate attack experiment. Longer perturb box length or shorter perturb-target distance cause success rates to significantly increase for all model and attack combinations, except for perturb box length in untargeted attack on Cascade R-CNN. The interaction terms, even when significant, are negligibly close to 0. Table headers are explained in Appendix ??.

Group		Regression						
Attack	term	sig	estimate	std.error	statistic	p.value	conf.low	conf.high
YOLOv3								
Vanishing	distance	*	-0.015	0.002	-7.681	0.000	-0.018	-0.011
	length	*	0.030	0.002	19.637	0.000	0.027	0.033
	distance * length	*	0.000	0.000	-6.081	0.000	0.000	0.000

Mislabeling	distance	*	-0.015	0.002	-8.540	0.000	-0.018	-0.011
	length	*	0.019	0.001	16.603	0.000	0.016	0.021
	distance * length		0.000	0.000	-1.733	0.083	0.000	0.000
Untargeted	distance	*	-0.021	0.003	-7.440	0.000	-0.026	-0.015
	length	*	0.007	0.001	6.528	0.000	0.005	0.009
	distance * length		0.000	0.000	1.467	0.142	0.000	0.000
SSD								
Vanishing	distance	*	-0.015	0.002	-7.055	0.000	-0.019	-0.011
	length	*	0.024	0.001	17.747	0.000	0.021	0.027
	distance * length	*	0.000	0.000	-4.823	0.000	0.000	0.000
Mislabeling	distance	*	-0.018	0.002	-7.553	0.000	-0.023	-0.014
	length	*	0.020	0.001	15.991	0.000	0.017	0.022
	distance * length	*	0.000	0.000	-2.458	0.014	0.000	0.000
Untargeted	distance	*	-0.018	0.002	-7.742	0.000	-0.023	-0.014
	length	*	0.013	0.001	11.665	0.000	0.011	0.015
	distance * length		0.000	0.000	-0.873	0.383	0.000	0.000
RetinaNet								
Vanishing	distance	*	-0.045	0.006	-7.187	0.000	-0.058	-0.033
	length	*	0.016	0.001	10.614	0.000	0.013	0.019
	distance * length	*	0.000	0.000	-2.147	0.032	0.000	0.000
Mislabeling	distance	*	-0.031	0.007	-4.240	0.000	-0.047	-0.018
	length	*	0.008	0.002	4.541	0.000	0.005	0.012
	distance * length		0.000	0.000	-1.021	0.307	0.000	0.000
Untargeted	distance	*	-0.038	0.005	-7.446	0.000	-0.049	-0.029
	length	*	0.005	0.001	3.969	0.000	0.003	0.008
	distance * length	*	0.000	0.000	6.925	0.000	0.000	0.000
Faster R-CNN								
Vanishing	distance	*	-0.061	0.010	-6.407	0.000	-0.081	-0.044
	length	*	0.011	0.001	7.127	0.000	0.008	0.014
	distance * length		0.000	0.000	-0.490	0.624	0.000	0.000
Mislabeling	distance	*	-0.054	0.012	-4.664	0.000	-0.080	-0.034
	length	*	0.007	0.002	3.706	0.000	0.003	0.010
	distance * length		0.000	0.000	-0.717	0.473	0.000	0.000
Untargeted	distance	*	-0.044	0.006	-8.012	0.000	-0.056	-0.034
	length	*	0.005	0.001	3.676	0.000	0.002	0.007
	distance * length	*	0.000	0.000	6.889	0.000	0.000	0.000
Cascade R-CNN								
Vanishing	distance	*	-0.063	0.010	-6.579	0.000	-0.083	-0.046
	length	*	0.015	0.002	9.395	0.000	0.012	0.018
	distance * length		0.000	0.000	-1.003	0.316	0.000	0.000
Mislabeling	distance	*	-0.062	0.012	-5.240	0.000	-0.088	-0.041

	length	*	0.010	0.002	5.795	0.000	0.006	0.013
	distance * length		0.000	0.000	-0.122	0.903	0.000	0.000
Untargeted	distance	*	-0.061	0.008	-7.544	0.000	-0.079	-0.047
	length		0.002	0.001	1.498	0.134	-0.001	0.005
	distance * length	*	0.000	0.000	6.198	0.000	0.000	0.000

```
reg_mod <- reg_res$mod
```

```
newdata <- expand_grid(
  bbox_dist = linear_space(data$bbox_dist),
  bbox_len = unique(data$bbox_len)
) |>
  glimpse()
```

```
## Rows: 400
```

```
## Columns: 2
```

```
## $ bbox_dist <dbl> 10.00000, 10.00000, 10.00000, 10.00000, 11.91919, 11.91919, ~
```

```
## $ bbox_len <dbl> 100, 10, 200, 50, 100, 10, 200, 50, 100, 10, 200, 50, 100, 1~
```

```
# type.predict = "link" by default
```

```
# https://broom.tidymodels.org/reference/augment.glm.html
```

```
# https://stackoverflow.com/questions/14423325/confidence-intervals-for-predictions-from-logistic-regre
```

```
reg_pred <- reg_mod |>
```

```
  summarize(augment(mod, newdata = newdata, se_fit = TRUE)) |>
```

```
  mutate(success = plogis(.fitted), ul = plogis(.fitted + 1.96 * .se.fit), ll = plogis(.fitted - 1.96 * .se.fit))
```

```
  glimpse()
```

```
## `summarise()` has grouped output by 'model_name', 'loss_target'. You can
```

```
## override using the `.groups` argument.
```

```
## Rows: 6,000
```

```
## Columns: 9
```

```
## Groups: model_name, loss_target [15]
```

```
## $ model_name <ord> YOLOv3, YOLOv3, YOLOv3, YOLOv3, YOLOv3, YOLOv3, YOLOv3, YO~
```

```
## $ loss_target <ord> Vanishing, Vanishing, Vanishing, Vanishing, Vanishing, Van~
```

```
## $ bbox_dist <dbl> 10.00000, 10.00000, 10.00000, 10.00000, 11.91919, 11.91919~
```

```
## $ bbox_len <dbl> 100, 10, 200, 50, 100, 10, 200, 50, 100, 10, 200, 50, 100, ~
```

```
## $ .fitted <dbl> 1.1206648, -1.4569515, 3.9846829, -0.3113443, 1.0753688, --
```

```
## $ .se.fit <dbl> 0.08354473, 0.11366498, 0.19249889, 0.08011917, 0.08206361~
```

```
## $ success <dbl> 0.7541120, 0.1889340, 0.9817412, 0.4227867, 0.7456166, 0.1~
```

```
## $ ul <dbl> 0.7831999, 0.2254527, 0.9874075, 0.4614987, 0.7749042, 0.2~
```

```
## $ ll <dbl> 0.7225041, 0.1571306, 0.9735935, 0.3850003, 0.7139250, 0.1~
```

```
arb_cap <- glue("{bold_tex('A deliberate attack obfuscates intent with increased success for all models and attacks:')}")
```

```
arb_cap
```

```
## \textbf{A deliberate attack obfuscates intent with increased success for all models and attacks:} W
```

```
g <- success_expanded_data |> ggplot(aes(bbox_dist, success, color = bbox_len, group = bbox_len)) +
```

```
  stat_summary(fun.data = "mean_cl_boot") +
```

```
  facet_grid(cols = vars(model_name), rows = vars(loss_target))
```

```
# https://github.com/tidyverse/ggplot2/blob/ef00be7e2016e1259b4aef7f7c85651df123beff/R/geom-smooth.r#L1
```

```
g <- g + geom_ribbon(
```

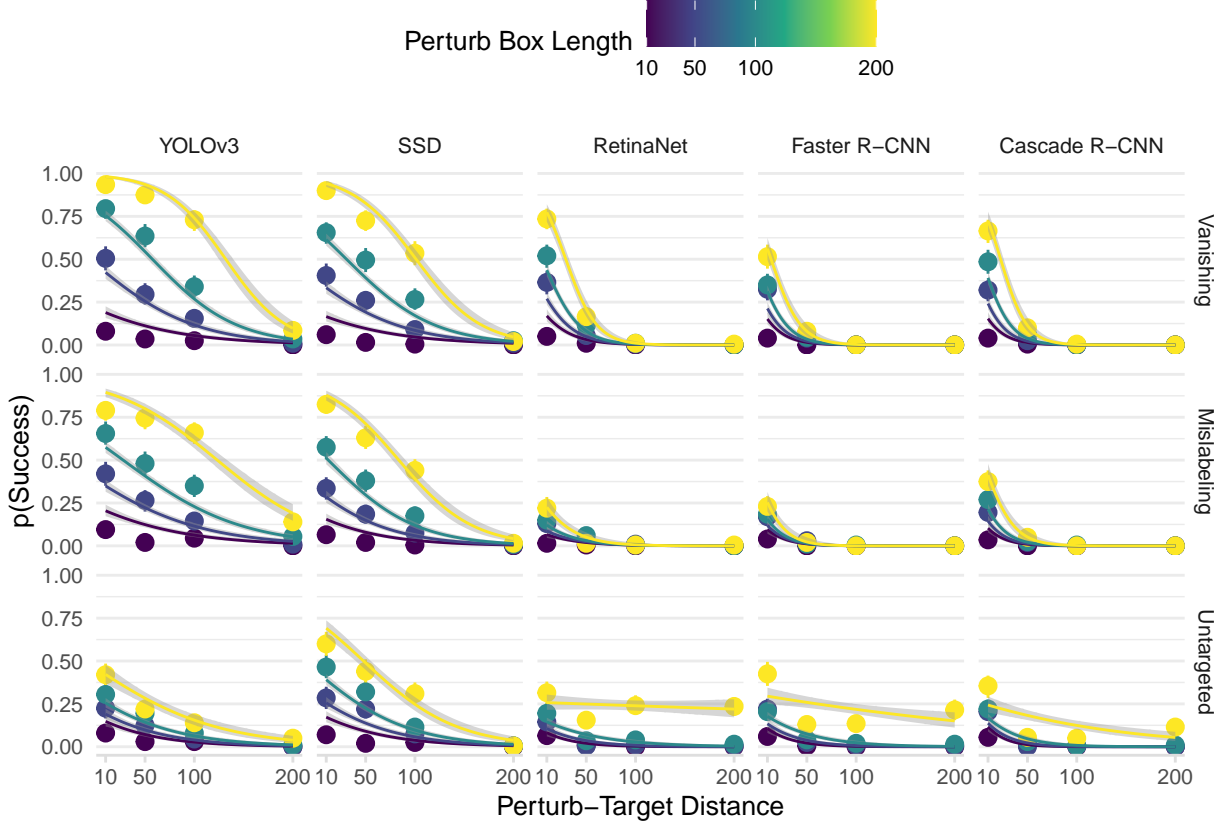


Figure 1: **A deliberate attack obfuscates intent with increased success for all models and attacks:** We implement intent obfuscating attack by perturbing an arbitrary non-overlapping square region to disrupt a randomly selected target object at various lengths and distances. The binned summaries and regression trendlines graph success proportion against perturb-target distance and perturb box length in the deliberate attack experiment. Errors are 95% confidence intervals. and every point aggregates success over 200 images. The deliberate attack multiplies success as compared to the randomized attack (Figure ??), especially at close perturb-target distance and large perturb box length. Full details are given in Section ??.

```
data = reg_pred, aes(ymin = ll, ymax = ul),
fill = "grey60", linetype = 0, alpha = 0.4
) +
geom_line(data = reg_pred)

g + labs(x = dist_lab, y = "p(Success)") +
  scale_x_continuous(breaks = unique(success_expanded_data$bbox_dist)) +
  scale_color_viridis_c(name = len_lab, breaks = unique(success_expanded_data$bbox_len))

get_reg_vars <- function(data) {
  data |> select(bbox_dist, bbox_size_perturb, model_name, loss_target, success, object)
}

# run random.Rmd 1st
rand_dist_size <- readRDS("./analysis/rand_dist_size.RDS") |>
  mutate(object = 1) |>
  get_reg_vars() |>
  glimpse()
```



```
## Rows: 75,000
## Columns: 6
## $ bbox_dist      <dbl> 11.250000, 74.020000, 267.290000, 161.231650, 61.260~
## $ bbox_size_perturb <dbl> 4705.2345, 3803.1889, 595.2576, 29362.0050, 43664.54~
## $ model_name      <ord> Cascade R-CNN, Cascade R-CNN, Cascade R-CNN, Cascade~
## $ loss_target     <ord> Mislabeling, Mislabeling, Mislabeling, Mislabeling, ~
## $ success         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~
## $ object          <dbl> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1~
```

```
comb_dist_size <- success_expanded_data |>
  mutate(object = 0, bbox_size_perturb = bbox_len^2) |>
  get_reg_vars() |>
  bind_rows(rand_dist_size) |>
  mutate(
    bbox_dist = bbox_dist / 1e2,
    bbox_size_perturb = bbox_size_perturb / 1e5
  ) |>
  glimpse()
```

```
## Rows: 123,000
## Columns: 6
## $ bbox_dist      <dbl> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1~
## $ bbox_size_perturb <dbl> 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.~
## $ model_name      <ord> Cascade R-CNN, Cascade R-CNN, Cascade R-CNN, Cascade~
## $ loss_target     <ord> Mislabeling, Mislabeling, Mislabeling, Mislabeling, ~
## $ success         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~
## $ object          <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0~
```

```
stopifnot(nrow(comb_dist_size) == nrow(success_expanded_data) +
  nrow(rand_dist_size) && sum(is.na(comb_dist_size)) == 0)
```

```
# control both
```

```
model <- partial(glm_model, predictor = "object + bbox_dist * bbox_size_perturb")
data <- comb_dist_size
```

```
reg_est <- get_tidied_reg(model, data)
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
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## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
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```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```


[illegible]

[illegible]

[illegible]

[illegible]

[illegible]


```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## `summarise()` has grouped output by 'model_name', 'loss_target'. You can
## override using the `.groups` argument.
```

```
ext_sig(reg_est, "both", "object")
```

```
## -----object-----
## Total 15 predictors:
## 10 (67%) significant;
## 10 (67%) both

## # A tibble: 10 x 9
## # Groups:   model_name, loss_target [10]
##   model_name loss_target term estimate std.error statistic p.value conf.low conf.high
##   <ord>      <ord>      <chr>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
## 1 YOLOv3      Vanishing obje~ -0.317    0.063    -5.03      0      -0.44    -0.193
## 2 YOLOv3      Untarget~ obje~ -0.201    0.079    -2.54     0.011   -0.357   -0.046
## 3 SSD          Untarget~ obje~  0.176    0.066     2.66     0.008    0.047    0.306
## 4 RetinaNet    Vanishing obje~ -0.551    0.093    -5.95      0      -0.734   -0.37
## 5 RetinaNet    Untarget~ obje~ -0.448    0.091    -4.90      0      -0.628   -0.269
## 6 Faster R-CNN Vanishing obje~ -0.768    0.113    -6.81      0      -0.991   -0.549
## 7 Faster R-CNN Mislabel~ obje~ -0.384    0.139    -2.77     0.006   -0.657   -0.113
## 8 Faster R-CNN Untarget~ obje~ -0.275    0.089    -3.10     0.002   -0.449   -0.101
## 9 Cascade R-CNN Vanishing obje~ -0.665    0.104    -6.40      0      -0.87    -0.462
## 10 Cascade R-CNN Mislabel~ obje~ -0.282    0.117    -2.40     0.016   -0.513   -0.052
## # ... with abbreviated variable names 1: loss_target, 2: estimate,
## # 3: std.error, 4: statistic, 5: conf.low, 6: conf.high
```

```
ext_sig(reg_est, "neg", "bbox_dist")
```

```
## -----bbox_dist-----
## Total 15 predictors:
## 15 (100%) significant;
## 15 (100%) neg

## # A tibble: 15 x 9
## # Groups:   model_name, loss_target [15]
##   model_name loss_target term estimate std.error statistic p.value conf.low conf.high
##   <ord>      <ord>      <chr>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
## 1 YOLOv3      Vanishing bbox~ -1.71    0.076   -22.4      0      -1.86   -1.56
## 2 YOLOv3      Mislabel~ bbox~ -1.52    0.066   -22.8      0      -1.65   -1.39
## 3 YOLOv3      Untarget~ bbox~ -1.97    0.114   -17.4      0      -2.20   -1.75
## 4 SSD          Vanishing bbox~ -1.92    0.084   -23.0      0      -2.09   -1.76
## 5 SSD          Mislabel~ bbox~ -1.95    0.091   -21.4      0      -2.13   -1.78
## 6 SSD          Untarget~ bbox~ -2.06    0.093   -22.0      0      -2.25   -1.88
```

```
## 7 RetinaNet Vanishing bbox~ -4.95 0.261 -19.0 0 -5.47 -4.45
## 8 RetinaNet Mislabel~ bbox~ -3.97 0.357 -11.1 0 -4.70 -3.30
## 9 RetinaNet Untarget~ bbox~ -1.33 0.106 -12.6 0 -1.55 -1.13
## 10 Faster R-CNN Vanishing bbox~ -6.00 0.408 -14.7 0 -6.83 -5.23
## 11 Faster R-CNN Mislabel~ bbox~ -5.87 0.483 -12.1 0 -6.86 -4.96
## 12 Faster R-CNN Untarget~ bbox~ -1.80 0.124 -14.6 0 -2.05 -1.57
## 13 Cascade R-CNN Vanishing bbox~ -6.50 0.388 -16.7 0 -7.28 -5.76
## 14 Cascade R-CNN Mislabel~ bbox~ -6.32 0.438 -14.4 0 -7.21 -5.49
## 15 Cascade R-CNN Untarget~ bbox~ -2.46 0.159 -15.5 0 -2.79 -2.16
## # ... with abbreviated variable names 1: loss_target, 2: estimate,
## # 3: std.error, 4: statistic, 5: conf.low, 6: conf.high
```

```
ext_sig(reg_est, "pos", "bbox_size_perturb")
```

```
## -----bbox_size_perturb-----
## Total 15 predictors:
## 14 (93%) significant;
## 14 (93%) pos

## # A tibble: 14 x 9
## # Groups:   model_name, loss_target [14]
##   model_name loss_ta~1 term estim~2 std.e~3 stati~4 p.value conf.~5 conf.~6
##   <ord>      <ord>    <chr>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
## 1 YOLOv3     Vanishing bbox~ 8.59 0.367 23.4 0 7.88 9.32
## 2 YOLOv3     Mislabel~ bbox~ 4.54 0.253 17.9 0 4.05 5.04
## 3 YOLOv3     Untarget~ bbox~ 1.59 0.17 9.36 0 1.26 1.93
## 4 SSD        Vanishing bbox~ 5.88 0.296 19.9 0 5.31 6.47
## 5 SSD        Mislabel~ bbox~ 4.23 0.249 17.0 0 3.75 4.73
## 6 SSD        Untarget~ bbox~ 1.96 0.187 10.5 0 1.60 2.33
## 7 RetinaNet  Vanishing bbox~ 2.69 0.251 10.7 0 2.21 3.19
## 8 RetinaNet  Mislabel~ bbox~ 1.16 0.231 5.03 0 0.712 1.62
## 9 RetinaNet  Untarget~ bbox~ 1.68 0.165 10.2 0 1.36 2.00
## 10 Faster R-CNN Vanishing bbox~ 2.06 0.256 8.05 0 1.57 2.58
## 11 Faster R-CNN Untarget~ bbox~ 2.10 0.182 11.6 0 1.75 2.46
## 12 Cascade R-CNN Vanishing bbox~ 2.90 0.277 10.5 0 2.38 3.46
## 13 Cascade R-CNN Mislabel~ bbox~ 0.886 0.22 4.02 0 0.459 1.32
## 14 Cascade R-CNN Untarget~ bbox~ 0.913 0.161 5.68 0 0.598 1.23
## # ... with abbreviated variable names 1: loss_target, 2: estimate,
## # 3: std.error, 4: statistic, 5: conf.low, 6: conf.high
```

```
ext_sig(reg_est, "both", "bbox_dist:bbox_size_perturb")
```

```
## -----bbox_dist:bbox_size_perturb-----
## Total 15 predictors:
## 8 (53%) significant;
## 8 (53%) both

## # A tibble: 8 x 9
## # Groups:   model_name, loss_target [8]
##   model_name loss_tar~1 term estim~2 std.e~3 stati~4 p.value conf.~5 conf.~6
##   <ord>      <ord>    <chr>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
## 1 YOLOv3     Vanishing bbox~ -3.26 0.31 -10.5 0 -3.87 -2.66
## 2 YOLOv3     Mislabeli~ bbox~ -0.908 0.231 -3.94 0 -1.36 -0.462
## 3 SSD        Vanishing bbox~ -2.26 0.289 -7.83 0 -2.84 -1.71
## 4 SSD        Mislabeli~ bbox~ -1.51 0.282 -5.36 0 -2.08 -0.971
## 5 RetinaNet  Untargeted bbox~ 1.77 0.203 8.70 0 1.37 2.17
```

```
## 6 Faster R-CNN Mislabeli~ bbox~ 2.06 0.747 2.75 0.006 0.44 3.36
## 7 Faster R-CNN Untargeted bbox~ 2.23 0.233 9.57 0 1.78 2.69
## 8 Cascade R-CNN Untargeted bbox~ 2.09 0.216 9.69 0 1.67 2.52
## # ... with abbreviated variable names 1: loss_target, 2: estimate,
## # 3: std.error, 4: statistic, 5: conf.low, 6: conf.high

dist_lab <- "Perturb-Target Distance (100 pixels)"
size_lab <- "Perturb Box Size (100,000 squared pixels)"

pred_name <- glue("object (versus non-object), with {dist_lab} and {size_lab} as covariates")
main_pt <- "perturbing an object (in the randomized attack) rather than a non-object (in the deliberate attack)"

tab_cap <- glue("We combined the data in the randomized and deliberate attack experiments to run a logistic regression")

print_statistics(reg_est, tab_cap)
```

Table 2: We combined the data in the randomized and deliberate attack experiments to run a logistic model regressing success against object (versus non-object), with perturb-target distance (100 pixels) and perturb box size (100,000 squared pixels) as covariates. The “object” term codes object as 1 and non-object as 0. Perturbing an object (in the randomized attack) rather than a non-object (in the deliberate attack) significantly decreases success rates for most model and attack combinations, after controlling for perturb sizes and perturb-target distances. Table headers are explained in Appendix ??.

Group		Regression						
Attack	term	sig	estimate	std.error	statistic	p.value	conf.low	conf.high
YOLOv3								
Vanishing	object	*	-0.317	0.063	-5.031	0.000	-0.440	-0.193
	distance	*	-1.711	0.076	-22.383	0.000	-1.863	-1.563
	size	*	8.585	0.367	23.423	0.000	7.878	9.315
	distance * size	*	-3.258	0.310	-10.498	0.000	-3.872	-2.655
Mislabeling	object		-0.026	0.059	-0.440	0.660	-0.141	0.089
	distance	*	-1.515	0.066	-22.796	0.000	-1.647	-1.386
	size	*	4.538	0.253	17.940	0.000	4.050	5.041
	distance * size	*	-0.908	0.231	-3.938	0.000	-1.365	-0.462
Untargeted	object	*	-0.201	0.079	-2.544	0.011	-0.357	-0.046
	distance	*	-1.970	0.114	-17.351	0.000	-2.197	-1.752
	size	*	1.593	0.170	9.364	0.000	1.265	1.932
	distance * size		0.356	0.250	1.423	0.155	-0.149	0.837
SSD								
Vanishing	object		0.096	0.063	1.532	0.125	-0.027	0.219
	distance	*	-1.924	0.084	-22.955	0.000	-2.090	-1.762
	size	*	5.883	0.296	19.896	0.000	5.313	6.472
	distance * size	*	-2.263	0.289	-7.826	0.000	-2.844	-1.707
Mislabeling	object		-0.039	0.064	-0.609	0.542	-0.166	0.087
	distance	*	-1.953	0.091	-21.407	0.000	-2.134	-1.776
	size	*	4.228	0.249	16.958	0.000	3.749	4.726

		distance * size *	-1.509	0.282	-5.356	0.000	-2.077	-0.971
Untargeted	object	*	0.176	0.066	2.661	0.008	0.047	0.306
	distance	*	-2.060	0.093	-22.041	0.000	-2.246	-1.879
	size	*	1.958	0.187	10.482	0.000	1.599	2.331
	distance * size		-0.227	0.244	-0.929	0.353	-0.719	0.240
RetinaNet								
Vanishing	object	*	-0.551	0.093	-5.947	0.000	-0.734	-0.370
	distance	*	-4.949	0.261	-18.960	0.000	-5.472	-4.448
	size	*	2.686	0.251	10.722	0.000	2.208	3.190
	distance * size		-0.881	0.569	-1.548	0.122	-2.035	0.197
Mislabeling	object		-0.245	0.136	-1.799	0.072	-0.513	0.022
	distance	*	-3.968	0.357	-11.109	0.000	-4.697	-3.297
	size	*	1.163	0.231	5.032	0.000	0.712	1.621
	distance * size		0.117	0.696	0.168	0.867	-1.323	1.403
Untargeted	object	*	-0.448	0.091	-4.902	0.000	-0.628	-0.269
	distance	*	-1.333	0.106	-12.560	0.000	-1.546	-1.130
	size	*	1.675	0.165	10.157	0.000	1.355	2.002
	distance * size	*	1.766	0.203	8.701	0.000	1.373	2.170
Faster R-CNN								
Vanishing	object	*	-0.768	0.113	-6.813	0.000	-0.991	-0.549
	distance	*	-6.002	0.408	-14.728	0.000	-6.829	-5.230
	size	*	2.062	0.256	8.052	0.000	1.572	2.577
	distance * size		-1.190	0.905	-1.315	0.188	-3.059	0.485
Mislabeling	object	*	-0.384	0.139	-2.770	0.006	-0.657	-0.113
	distance	*	-5.868	0.483	-12.144	0.000	-6.858	-4.961
	size		0.461	0.252	1.832	0.067	-0.029	0.958
	distance * size	*	2.055	0.747	2.752	0.006	0.440	3.362
Untargeted	object	*	-0.275	0.089	-3.096	0.002	-0.449	-0.101
	distance	*	-1.804	0.124	-14.599	0.000	-2.053	-1.568
	size	*	2.104	0.182	11.585	0.000	1.752	2.464
	distance * size	*	2.226	0.233	9.570	0.000	1.778	2.690
Cascade R-CNN								
Vanishing	object	*	-0.665	0.104	-6.395	0.000	-0.870	-0.462
	distance	*	-6.496	0.388	-16.731	0.000	-7.279	-5.757
	size	*	2.905	0.277	10.474	0.000	2.378	3.465
	distance * size		-1.579	0.840	-1.881	0.060	-3.310	-0.020
Mislabeling	object	*	-0.282	0.117	-2.402	0.016	-0.513	-0.052
	distance	*	-6.317	0.438	-14.410	0.000	-7.210	-5.489
	size	*	0.886	0.220	4.018	0.000	0.459	1.325
	distance * size		1.310	0.746	1.757	0.079	-0.265	2.666
Untargeted	object		-0.175	0.100	-1.739	0.082	-0.371	0.022

distance	*	-2.464	0.159	-15.457	0.000	-2.786	-2.160
size	*	0.913	0.161	5.677	0.000	0.598	1.229
distance * size	*	2.093	0.216	9.686	0.000	1.670	2.519
