

Plotting all the results

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2/19/2022

Jimmy needs to give me the right estimate of effect right now here is what I am using

```
smaller_true_ATE <- 0.15
bigger_true_ATE <- 0.3

pos_beta <- 1
neg_beta <- -1
```

Loading Data

Compiling Binary Data

Get all the odd numbers $\beta_1 = 0.767$

```
binary_final_odd <-
  binary_scen_1 %>%
    mutate(n_sample = 1000, beta1 = 0.767, desired_prop = 0.1) %>%
  bind_rows(binary_scen_3 %>%
    mutate(n_sample = 1000, beta1 = 0.767, desired_prop = 0.2)) %>%
  bind_rows(binary_scen_5 %>%
    mutate(n_sample = 1000, beta1 = 0.767, desired_prop = 0.3)) %>%
  bind_rows(binary_scen_13 %>%
    mutate(n_sample = 100, beta1 = 0.767, desired_prop = 0.1)) %>%
  bind_rows(binary_scen_15 %>%
    mutate(n_sample = 100, beta1 = 0.767, desired_prop = 0.2)) %>%
  bind_rows(binary_scen_17 %>%
    mutate(n_sample = 100, beta1 = 0.767, desired_prop = 0.3))

binary_final_odd <- binary_final_odd %>%
  mutate(
    ATE_bias = ATE - smaller_true_ATE,
    empirical_bias = empirical_mean - smaller_true_ATE,
    boot_type = ifelse(boot_type == 0, "Simple", "Complex")
  )

rm(binary_scen_1, binary_scen_3, binary_scen_5,
  binary_scen_13, binary_scen_15, binary_scen_17)
```

Get all the even numbers $\beta_1 = 1.386$

```

binary_final_even <-
  binary_scen_2 %>%
    mutate(n_sample = 1000, beta1 = 1.386, desired_prop = 0.1) %>%
  bind_rows(binary_scen_4 %>%
    mutate(n_sample = 1000, beta1 = 1.386, desired_prop = 0.2)) %>%
  bind_rows(binary_scen_6 %>%
    mutate(n_sample = 1000, beta1 = 1.386, desired_prop = 0.3)) %>%
  bind_rows(binary_scen_14 %>%
    mutate(n_sample = 100, beta1 = 1.386, desired_prop = 0.1)) %>%
  bind_rows(binary_scen_16 %>%
    mutate(n_sample = 100, beta1 = 1.386, desired_prop = 0.2)) %>%
  bind_rows(binary_scen_18 %>%
    mutate(n_sample = 100, beta1 = 1.386, desired_prop = 0.3))

binary_final_even <- binary_final_even %>%
  mutate(
    ATE_bias = ATE - bigger_true_ATE,
    empirical_bias = empirical_mean - bigger_true_ATE,
    boot_type = ifelse(boot_type == 0, "Simple", "Complex")
  )

rm(binary_scen_2, binary_scen_4, binary_scen_6,
  binary_scen_14, binary_scen_16, binary_scen_18)

```

```

binary_final <- binary_final_even %>% bind_rows(binary_final_odd)

```

Compiling Continuous Data

```

continuous_final_odd <-
  cont_df_scen_1 %>%
    mutate(n_sample = 1000, beta1 = pos_beta, desired_prop = 0.1) %>%
  bind_rows(cont_df_scen_3 %>%
    mutate(n_sample = 1000, beta1 = pos_beta, desired_prop = 0.2)) %>%
  bind_rows(cont_df_scen_5 %>%
    mutate(n_sample = 1000, beta1 = pos_beta, desired_prop = 0.3)) %>%
  bind_rows(cont_df_scen_13 %>%
    mutate(n_sample = 100, beta1 = pos_beta, desired_prop = 0.1)) %>%
  bind_rows(cont_df_scen_15 %>%
    mutate(n_sample = 100, beta1 = pos_beta, desired_prop = 0.2)) %>%
  bind_rows(cont_df_scen_17 %>%
    mutate(n_sample = 100, beta1 = pos_beta, desired_prop = 0.3)) %>%
  mutate(
    ATE_bias = ATE - pos_beta,
    empirical_bias = empirical_mean - pos_beta,
    boot_type = ifelse(boot_type == 0, "Simple", "Complex")
  )

rm(cont_df_scen_1, cont_df_scen_3, cont_df_scen_5,
  cont_df_scen_13, cont_df_scen_15, cont_df_scen_17)

```

```

continuous_final_even <-
  cont_df_scen_2 %>%
    mutate(n_sample = 1000, beta1 = neg_beta, desired_prop = 0.1) %>%
  bind_rows(cont_df_scen_4 %>%
    mutate(n_sample = 1000, beta1 = neg_beta, desired_prop = 0.2)) %>%
  bind_rows(cont_df_scen_6 %>%
    mutate(n_sample = 1000, beta1 = neg_beta, desired_prop = 0.3)) %>%
  bind_rows(cont_df_scen_14 %>%
    mutate(n_sample = 100, beta1 = neg_beta, desired_prop = 0.1)) %>%
  bind_rows(cont_df_scen_16 %>%
    mutate(n_sample = 100, beta1 = neg_beta, desired_prop = 0.2)) %>%
  bind_rows(cont_df_scen_18 %>%
    mutate(n_sample = 100, beta1 = neg_beta, desired_prop = 0.3)) %>%
  mutate(
    ATE_bias = ATE - neg_beta,
    empirical_bias = empirical_mean - neg_beta,
    boot_type = ifelse(boot_type == 0, "Simple", "Complex")
  )

rm(cont_df_scen_2, cont_df_scen_4, cont_df_scen_6,
  cont_df_scen_14, cont_df_scen_16, cont_df_scen_18)

```

```

continuous_final <-
  continuous_final_odd %>%
  bind_rows(continuous_final_even)

rm(continuous_final_even, continuous_final_odd)

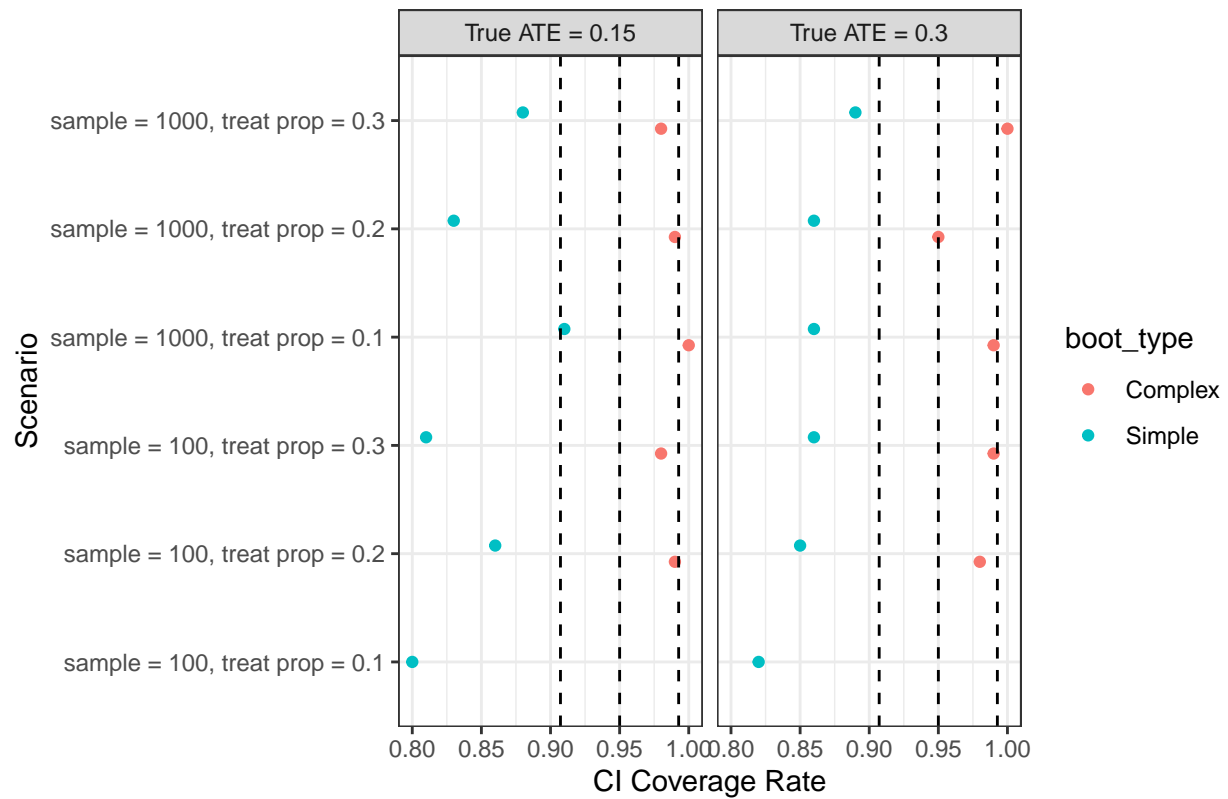
```

Binary Coverage Rates

'summarise()' has grouped output by 'new_name', 'treat_effect'. You can override
using the '.groups' argument.

name the scenarios by sample size and treat prop and facet by the treatment effect

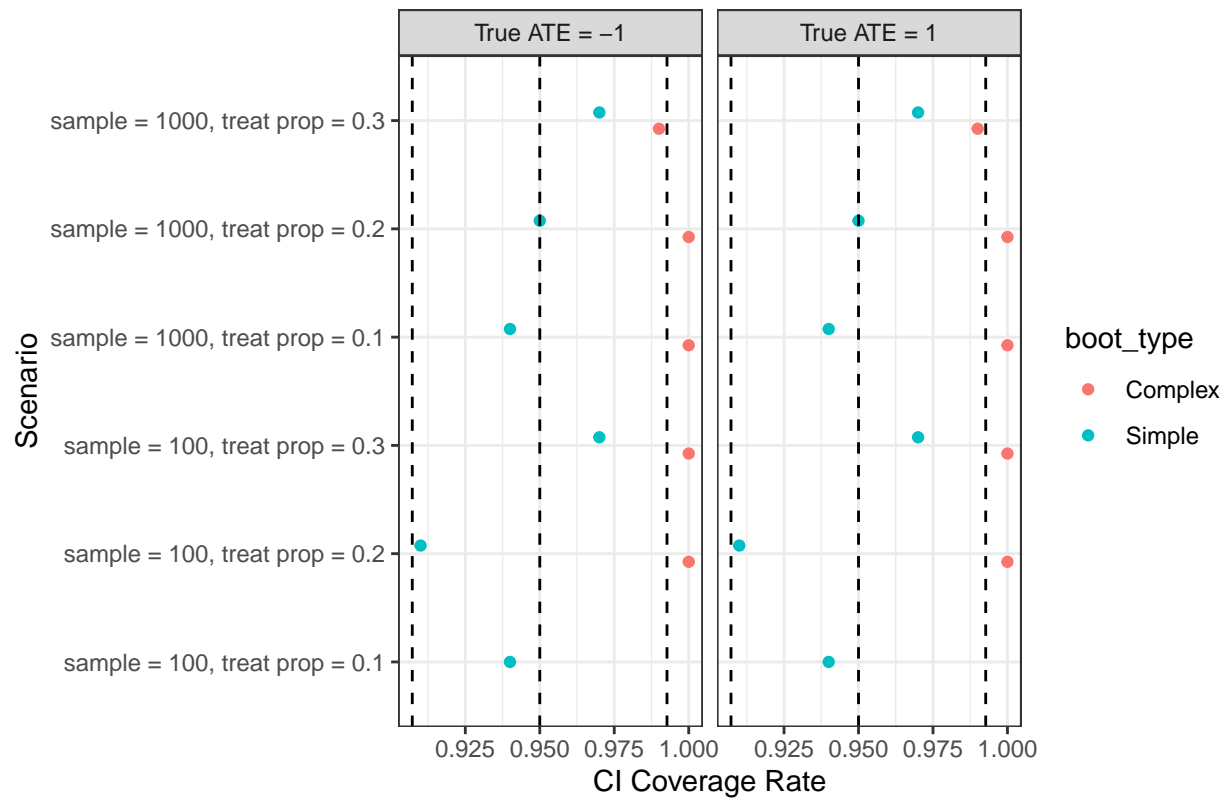
Binary Coverage Rates by Parameters of Interest



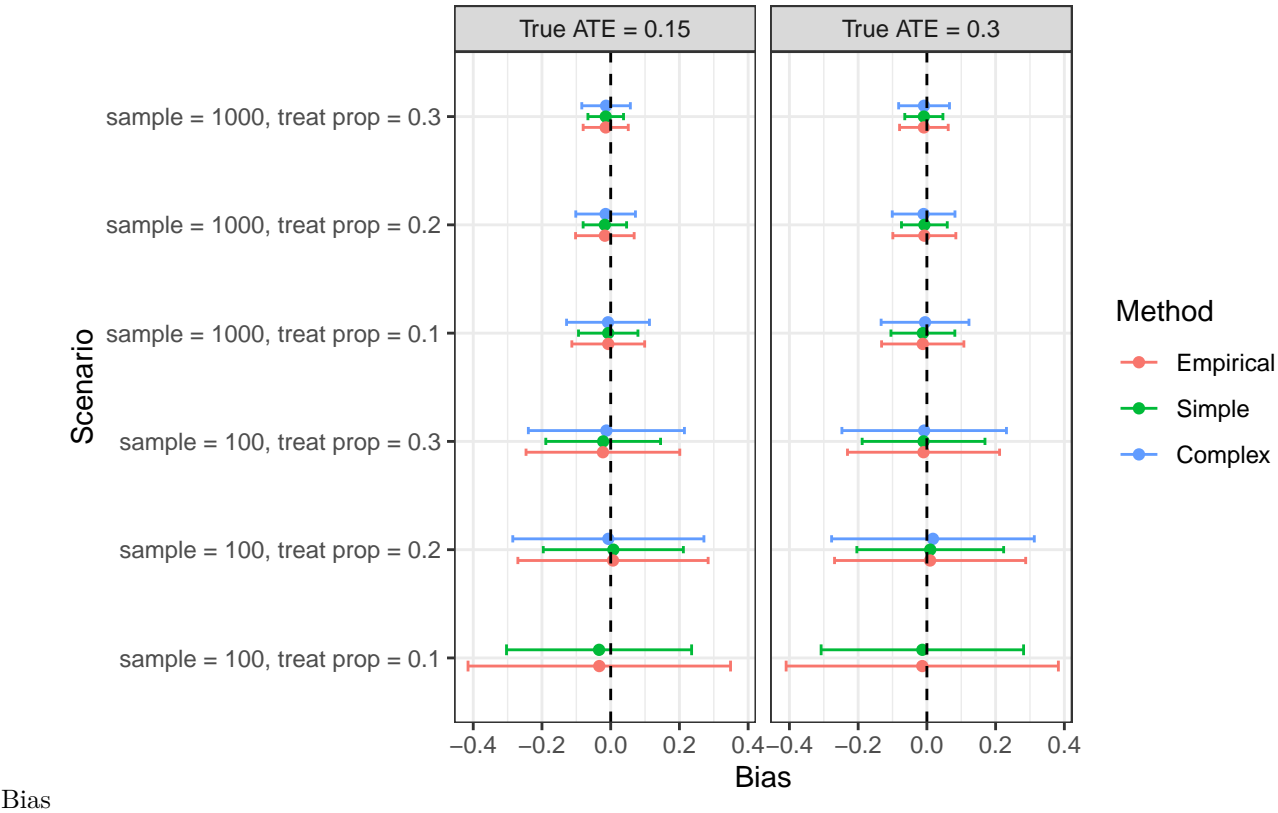
Continuous Coverage Rates

'summarise()' has grouped output by 'new_name', 'treat_effect'. You can override
using the '.groups' argument.

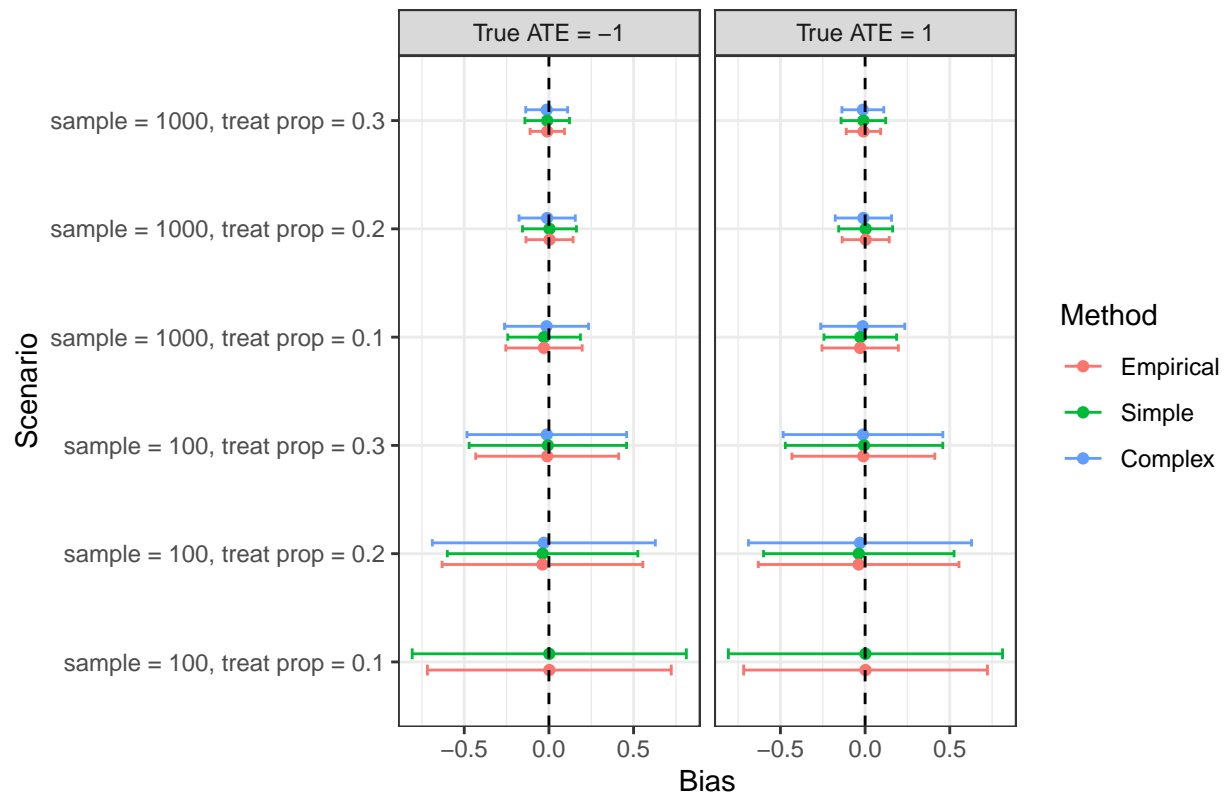
Continuous Coverage Rates by Parameters of Interest



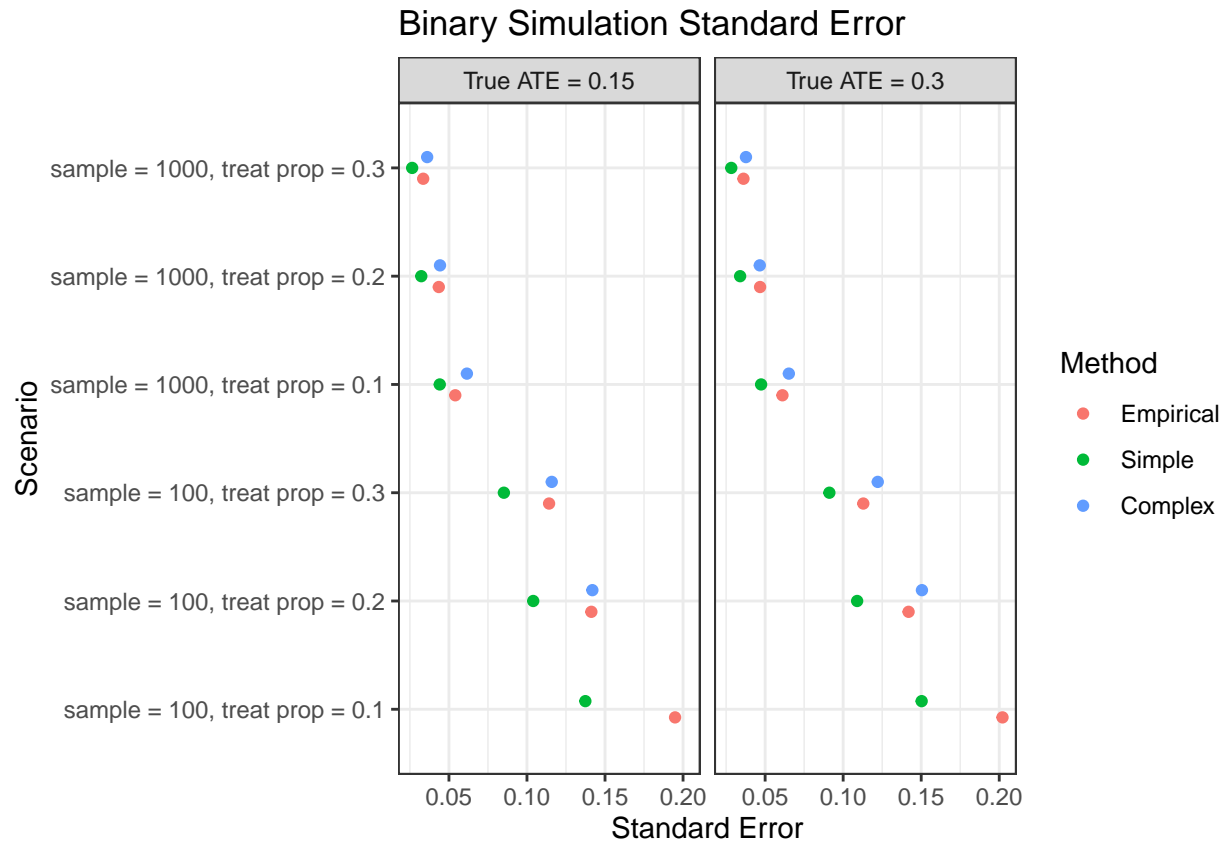
Binary Simulation Bias and Standard Error CI



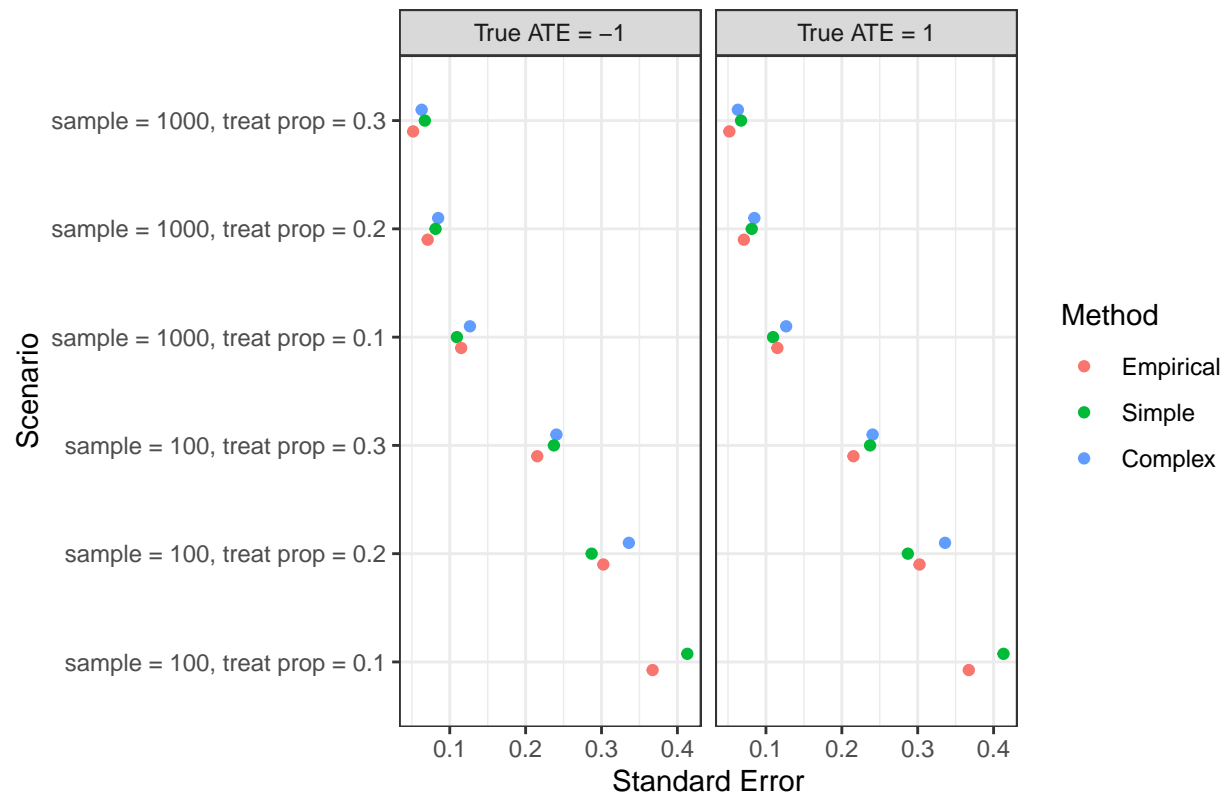
Continuous Simulation Bias and Standard Error CI



Standard Error



Continuous Simulation Standard Error



Binary Outcome	Empirical			Simple		Complex		
Scenario	E_SE	E_Bias	S_SE	S_Bias	S_CR	C_SE	C_Bias	C_CR
Large Sample, ATE = 0.15, p = 0.1	0.115	-0.029	0.054	-0.007	0.91	0.036	-0.008	1.00
Large Sample, ATE = 0.15, p = 0.2	0.071	0.004	0.043	-0.017	0.83	0.030	-0.015	0.99
Large Sample, ATE = 0.15, p = 0.3	0.052	-0.010	0.033	-0.014	0.88	0.028	-0.013	0.98
Large Sample, ATE = 0.30, p = 0.1	0.115	-0.029	0.061	-0.012	0.86	0.052	-0.005	0.99
Large Sample, ATE = 0.30, p = 0.2	0.071	0.004	0.047	-0.007	0.86	0.042	-0.010	0.95
Large Sample, ATE = 0.30, p = 0.3	0.052	-0.010	0.036	-0.009	0.89	0.032	-0.008	1.00
Small Sample, ATE = 0.15, p = 0.1	0.367	0.003	0.194	-0.034	0.80	NA	NA	NA
Small Sample, ATE = 0.15, p = 0.2	0.302	-0.038	0.140	0.008	0.86	0.109	-0.007	0.99
Small Sample, ATE = 0.15, p = 0.3	0.215	-0.010	0.115	-0.022	0.81	0.103	-0.013	0.98
Small Sample, ATE = 0.30, p = 0.1	0.367	0.003	0.202	-0.013	0.82	NA	NA	NA
Small Sample, ATE = 0.30, p = 0.2	0.302	-0.038	0.142	0.010	0.85	0.112	0.018	0.98
Small Sample, ATE = 0.30, p = 0.3	0.215	-0.010	0.113	-0.010	0.86	0.099	-0.008	0.99

Continuous Outcome	Empirical			Simple		Complex		
Scenario	E_SE	E_Bias	S_SE	S_Bias	S_CR	C_SE	C_Bias	C_CR
Large Sample, ATE = +1, p = 0.1	0.115	-0.029	0.114	-0.028	0.94	0.054	-0.014	1.00
Large Sample, ATE = +1, p = 0.2	0.071	0.004	0.071	0.003	0.95	0.041	-0.010	1.00
Large Sample, ATE = +1, p = 0.3	0.052	-0.010	0.052	-0.010	0.97	0.034	-0.013	0.99
Large Sample, ATE = -1, p = 0.1	0.115	-0.029	0.114	-0.028	0.94	0.054	-0.014	1.00
Large Sample, ATE = -1, p = 0.2	0.071	0.004	0.071	0.003	0.95	0.041	-0.010	1.00
Large Sample, ATE = -1, p = 0.3	0.052	-0.010	0.052	-0.010	0.97	0.034	-0.013	0.99
Small Sample, ATE = +1, p = 0.1	0.367	0.003	0.367	0.002	0.94	NA	NA	NA
Small Sample, ATE = +1, p = 0.2	0.302	-0.038	0.302	-0.037	0.91	0.138	-0.030	1.00
Small Sample, ATE = +1, p = 0.3	0.215	-0.010	0.218	-0.006	0.97	0.115	-0.012	1.00
Small Sample, ATE = -1, p = 0.1	0.367	0.003	0.367	0.002	0.94	NA	NA	NA
Small Sample, ATE = -1, p = 0.2	0.302	-0.038	0.302	-0.037	0.91	0.138	-0.030	1.00
Small Sample, ATE = -1, p = 0.3	0.215	-0.010	0.218	-0.006	0.97	0.115	-0.012	1.00