

Missing Data

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Varieties of Missingness

```
mc_listwise <- foreach(i = 1:1000, .combine = "rbind") %do% {  
  dat <- sim_data(100) %>%  
    mutate(x2 = ifelse(missing == 1, NA, x2))  
  cf <- do(dat %>% group_by(type),  
    lm(y ~ x1 + x2, data = .) %>% tidy())  
  cf  
}
```

```
mc_listwise %>%  
  group_by(type, term) %>%  
  summarise(mean = mean(estimate),  
    se_true = sd(estimate),  
    se_nominal = mean(std.error)) %>%  
  mutate_each(funs(round(., 2)), -type, -term)
```

```
## Source: local data frame [12 x 5]  
## Groups: type  
##  
##      type      term mean se_true se_nominal  
## 1 Complete (Intercept) 1.01    0.10    0.10  
## 2 Complete      x1 1.01    0.22    0.23  
## 3 Complete      x2 0.99    0.20    0.20  
## 4 MCAR (Intercept) 1.01    0.15    0.14  
## 5 MCAR      x1 1.02    0.33    0.33  
## 6 MCAR      x2 0.99    0.29    0.29  
## 7 MAR (Intercept) 1.01    0.18    0.18  
## 8 MAR      x1 1.01    0.33    0.35  
## 9 MAR      x2 0.99    0.29    0.30  
## 10 NI (Intercept) 1.00    0.19    0.18
```

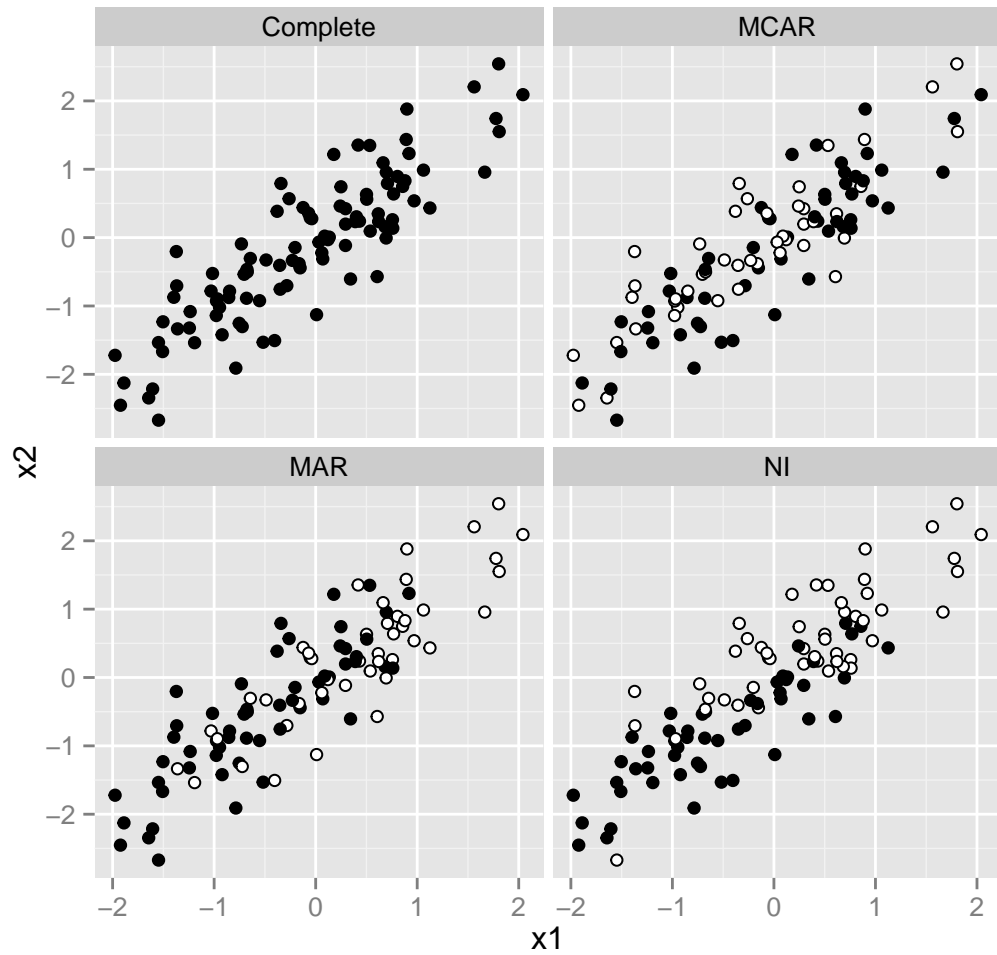


Figure 1. Illustration of data that are complete, missing completely at random (MCAR), missing at random (MAR), and nonignorably missing (NI).

```
## 11      NI      x1 1.01    0.33    0.33
## 12      NI      x2 0.99    0.32    0.31
```

```
mc_single <- foreach(i = 1:1000, .combine = "rbind") %do% {
  dat <- sim_data(100) %>%
    mutate(x2 = ifelse(missing == 1, NA, x2))

  cf <- do(dat %>% group_by(type), {
    x2_fit <- lm(x2 ~ x1, data = .)
    x2_pred <- predict(x2_fit, newdata = .)
    .$x2 <- ifelse(is.na(. $x2), x2_pred, . $x2)
    y_fit <- lm(y ~ x1 + x2, data = .)
    tidy(y_fit)
  })
}

mc_single %>%
  group_by(type, term) %>%
  summarise(mean = mean(estimate),
            se_true = sd(estimate),
            se_nominal = mean(std.error)) %>%
  mutate_each(funs(round(., 2)), -type, -term)
```

```
## Source: local data frame [12 x 5]
## Groups: type
##
##      type      term mean se_true se_nominal
## 1 Complete (Intercept) 1.00    0.10    0.10
## 2 Complete      x1 1.00    0.23    0.23
## 3 Complete      x2 1.00    0.20    0.20
## 4 MCAR (Intercept) 1.00    0.11    0.11
## 5 MCAR      x1 1.01    0.32    0.33
## 6 MCAR      x2 0.98    0.30    0.31
## 7 MAR (Intercept) 1.00    0.13    0.11
## 8 MAR      x1 0.99    0.33    0.33
## 9 MAR      x2 1.01    0.30    0.31
## 10 NI (Intercept) 1.18    0.14    0.12
## 11 NI      x1 1.09    0.31    0.32
## 12 NI      x2 1.01    0.31    0.33
```

References