

- We can use another loop to run `i` number of independent sample t-tests and to save the results of each test to a new data frame we are calling `result`

```
result <- NULL
```

```
for (i in 1:total_samples) {
  result <- rbind(tidy(t.test(filter(all_data, condition == "fast" & sample == i)$dv,
                                filter(all_data, condition == "slow" & sample == i)$dv,
                                paired = FALSE)), result)
}
```

```
> result
```

```
# A tibble: 100 x 10
```

| | estimate | estimate1 | estimate2 | statistic | p.value | parameter | conf.low | conf.high | method | alternative |
|----|----------|-----------|-----------|-----------|---------|-----------|----------|-----------|-------------------------|-------------|
| | <dbl> | <dbl> | <dbl> | <dbl> | <dbl> | <dbl> | <dbl> | <dbl> | <chr> | <chr> |
| 1 | 6.5 | 1013. | 1007. | 0.364 | 0.720 | 20.2 | -30.7 | 43.7 | Welch Two Sample t-test | two.sided |
| 2 | -15 | 1000. | 1015. | -0.695 | 0.494 | 22.0 | -59.7 | 29.7 | Welch Two Sample t-test | two.sided |
| 3 | 7.92 | 1019. | 1011. | 0.445 | 0.661 | 21.0 | -29.1 | 44.9 | Welch Two Sample t-test | two.sided |
| 4 | -16.4 | 984. | 1000. | -0.697 | 0.493 | 22.0 | -65.3 | 32.5 | Welch Two Sample t-test | two.sided |
| 5 | -10.8 | 1002. | 1012. | -0.517 | 0.612 | 16.9 | -54.7 | 33.2 | Welch Two Sample t-test | two.sided |
| 6 | 7.25 | 1000. | 993 | 0.359 | 0.723 | 20.4 | -34.8 | 49.3 | Welch Two Sample t-test | two.sided |
| 7 | -35 | 994. | 1030. | -1.66 | 0.113 | 20.6 | -79.0 | 8.99 | Welch Two Sample t-test | two.sided |
| 8 | -27.5 | 983 | 1010. | -1.40 | 0.175 | 21.8 | -68.2 | 13.2 | Welch Two Sample t-test | two.sided |
| 9 | 4.83 | 1026. | 1021. | 0.246 | 0.808 | 18.3 | -36.3 | 46.0 | Welch Two Sample t-test | two.sided |
| 10 | -35.8 | 996. | 1032 | -1.58 | 0.130 | 18.9 | -83.2 | 11.5 | Welch Two Sample t-test | two.sided |

```
# ... with 90 more rows
```

- We can work out for how many of the 100 tests we have found a significant difference at $< .05$ - and remember, there is actually a real difference (of 20 ms.) in the two population distributions we sampled from!

```
> count(filter(result, p.value < .05))  
# A tibble: 1 x 1  
      n  
  <int>  
1    17
```

- So, less than a fifth of the time are we finding a significant difference even though one exists in the distributions we sampled from. So with a sample size of 24 (12 per group) power to detect the effect we are looking for is .17