

HW 04

TUAN BUI

9/28/2021

Question 01

```
install.packages('car')
```

```
library(carData)
```

```
data("Salaries", package = "carData")
```

```
Salaries
```

| ## | rank | discipline | yrs.since.phd | yrs.service | sex | salary |
|-------|-----------|------------|---------------|-------------|--------|--------|
| ## 1 | Prof | B | 19 | 18 | Male | 139750 |
| ## 2 | Prof | B | 20 | 16 | Male | 173200 |
| ## 3 | AsstProf | B | 4 | 3 | Male | 79750 |
| ## 4 | Prof | B | 45 | 39 | Male | 115000 |
| ## 5 | Prof | B | 40 | 41 | Male | 141500 |
| ## 6 | AssocProf | B | 6 | 6 | Male | 97000 |
| ## 7 | Prof | B | 30 | 23 | Male | 175000 |
| ## 8 | Prof | B | 45 | 45 | Male | 147765 |
| ## 9 | Prof | B | 21 | 20 | Male | 119250 |
| ## 10 | Prof | B | 18 | 18 | Female | 129000 |
| ## 11 | AssocProf | B | 12 | 8 | Male | 119800 |
| ## 12 | AsstProf | B | 7 | 2 | Male | 79800 |
| ## 13 | AsstProf | B | 1 | 1 | Male | 77700 |
| ## 14 | AsstProf | B | 2 | 0 | Male | 78000 |
| ## 15 | Prof | B | 20 | 18 | Male | 104800 |
| ## 16 | Prof | B | 12 | 3 | Male | 117150 |
| ## 17 | Prof | B | 19 | 20 | Male | 101000 |
| ## 18 | Prof | A | 38 | 34 | Male | 103450 |
| ## 19 | Prof | A | 37 | 23 | Male | 124750 |
| ## 20 | Prof | A | 39 | 36 | Female | 137000 |
| ## 21 | Prof | A | 31 | 26 | Male | 89565 |
| ## 22 | Prof | A | 36 | 31 | Male | 102580 |
| ## 23 | Prof | A | 34 | 30 | Male | 93904 |
| ## 24 | Prof | A | 24 | 19 | Male | 113068 |
| ## 25 | AssocProf | A | 13 | 8 | Female | 74830 |
| ## 26 | Prof | A | 21 | 8 | Male | 106294 |
| ## 27 | Prof | A | 35 | 23 | Male | 134885 |
| ## 28 | AsstProf | B | 5 | 3 | Male | 82379 |
| ## 29 | AsstProf | B | 11 | 0 | Male | 77000 |
| ## 30 | Prof | B | 12 | 8 | Male | 118223 |
| ## 31 | Prof | B | 20 | 4 | Male | 132261 |
| ## 32 | AsstProf | B | 7 | 2 | Male | 79916 |
| ## 33 | Prof | B | 13 | 9 | Male | 117256 |

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|-------|-----------|---|----|----|--------|--------|
| ## 34 | AsstProf | B | 4 | 2 | Male | 80225 |
| ## 35 | AsstProf | B | 4 | 2 | Female | 80225 |
| ## 36 | AsstProf | B | 5 | 0 | Female | 77000 |
| ## 37 | Prof | B | 22 | 21 | Male | 155750 |
| ## 38 | AsstProf | B | 7 | 4 | Male | 86373 |
| ## 39 | Prof | B | 41 | 31 | Male | 125196 |
| ## 40 | AssocProf | B | 9 | 9 | Male | 100938 |
| ## 41 | Prof | B | 23 | 2 | Male | 146500 |
| ## 42 | AssocProf | B | 23 | 23 | Male | 93418 |
| ## 43 | Prof | B | 40 | 27 | Male | 101299 |
| ## 44 | Prof | B | 38 | 38 | Male | 231545 |
| ## 45 | Prof | B | 19 | 19 | Male | 94384 |
| ## 46 | Prof | B | 25 | 15 | Male | 114778 |
| ## 47 | Prof | B | 40 | 28 | Male | 98193 |
| ## 48 | Prof | B | 23 | 19 | Female | 151768 |
| ## 49 | Prof | B | 25 | 25 | Female | 140096 |
| ## 50 | AsstProf | B | 1 | 1 | Male | 70768 |
| ## 51 | Prof | B | 28 | 28 | Male | 126621 |
| ## 52 | Prof | B | 12 | 11 | Male | 108875 |
| ## 53 | AsstProf | B | 11 | 3 | Female | 74692 |
| ## 54 | Prof | B | 16 | 9 | Male | 106639 |
| ## 55 | AssocProf | B | 12 | 11 | Male | 103760 |
| ## 56 | AssocProf | B | 14 | 5 | Male | 83900 |
| ## 57 | Prof | B | 23 | 21 | Male | 117704 |
| ## 58 | AssocProf | B | 9 | 8 | Male | 90215 |
| ## 59 | AssocProf | B | 10 | 9 | Male | 100135 |
| ## 60 | AsstProf | B | 8 | 3 | Male | 75044 |
| ## 61 | AssocProf | B | 9 | 8 | Male | 90304 |
| ## 62 | AsstProf | B | 3 | 2 | Male | 75243 |
| ## 63 | Prof | B | 33 | 31 | Male | 109785 |
| ## 64 | AssocProf | B | 11 | 11 | Female | 103613 |
| ## 65 | AsstProf | B | 4 | 3 | Male | 68404 |
| ## 66 | AssocProf | B | 9 | 8 | Male | 100522 |
| ## 67 | Prof | B | 22 | 12 | Male | 101000 |
| ## 68 | Prof | B | 35 | 31 | Male | 99418 |
| ## 69 | Prof | B | 17 | 17 | Female | 111512 |
| ## 70 | Prof | B | 28 | 36 | Male | 91412 |
| ## 71 | Prof | B | 17 | 2 | Male | 126320 |
| ## 72 | Prof | B | 45 | 45 | Male | 146856 |
| ## 73 | Prof | B | 29 | 19 | Male | 100131 |
| ## 74 | Prof | B | 35 | 34 | Male | 92391 |
| ## 75 | Prof | B | 28 | 23 | Male | 113398 |
| ## 76 | AsstProf | B | 8 | 3 | Male | 73266 |
| ## 77 | Prof | B | 17 | 3 | Male | 150480 |
| ## 78 | Prof | B | 26 | 19 | Male | 193000 |
| ## 79 | AsstProf | B | 3 | 1 | Male | 86100 |
| ## 80 | AsstProf | B | 6 | 2 | Male | 84240 |
| ## 81 | Prof | B | 43 | 28 | Male | 150743 |
| ## 82 | Prof | B | 17 | 16 | Male | 135585 |
| ## 83 | Prof | B | 22 | 20 | Male | 144640 |

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|--------|-----------|---|----|----|--------|--------|
| ## 84 | AsstProf | B | 6 | 2 | Male | 88825 |
| ## 85 | Prof | B | 17 | 18 | Female | 122960 |
| ## 86 | Prof | B | 15 | 14 | Male | 132825 |
| ## 87 | Prof | B | 37 | 37 | Male | 152708 |
| ## 88 | AsstProf | B | 2 | 2 | Male | 88400 |
| ## 89 | Prof | B | 25 | 25 | Male | 172272 |
| ## 90 | AssocProf | B | 9 | 7 | Male | 107008 |
| ## 91 | AsstProf | B | 10 | 5 | Female | 97032 |
| ## 92 | AssocProf | B | 10 | 7 | Male | 105128 |
| ## 93 | AssocProf | B | 10 | 7 | Male | 105631 |
| ## 94 | Prof | B | 38 | 38 | Male | 166024 |
| ## 95 | Prof | B | 21 | 20 | Male | 123683 |
| ## 96 | AsstProf | B | 4 | 0 | Male | 84000 |
| ## 97 | AssocProf | B | 17 | 12 | Male | 95611 |
| ## 98 | Prof | B | 13 | 7 | Male | 129676 |
| ## 99 | Prof | B | 30 | 14 | Male | 102235 |
| ## 100 | Prof | B | 41 | 26 | Male | 106689 |
| ## 101 | Prof | B | 42 | 25 | Male | 133217 |
| ## 102 | Prof | B | 28 | 23 | Male | 126933 |
| ## 103 | Prof | B | 16 | 5 | Male | 153303 |
| ## 104 | Prof | B | 20 | 14 | Female | 127512 |
| ## 105 | AssocProf | A | 18 | 10 | Male | 83850 |
| ## 106 | Prof | A | 31 | 28 | Male | 113543 |
| ## 107 | AssocProf | A | 11 | 8 | Male | 82099 |
| ## 108 | AssocProf | A | 10 | 8 | Male | 82600 |
| ## 109 | AssocProf | A | 15 | 8 | Male | 81500 |
| ## 110 | Prof | A | 40 | 31 | Male | 131205 |
| ## 111 | Prof | A | 20 | 16 | Male | 112429 |
| ## 112 | AssocProf | A | 19 | 16 | Male | 82100 |
| ## 113 | AsstProf | A | 3 | 1 | Male | 72500 |
| ## 114 | Prof | A | 37 | 37 | Male | 104279 |
| ## 115 | Prof | A | 12 | 0 | Female | 105000 |
| ## 116 | Prof | A | 21 | 9 | Male | 120806 |
| ## 117 | Prof | A | 30 | 29 | Male | 148500 |
| ## 118 | Prof | A | 39 | 36 | Male | 117515 |
| ## 119 | AsstProf | A | 4 | 1 | Male | 72500 |
| ## 120 | AsstProf | A | 5 | 3 | Female | 73500 |
| ## 121 | Prof | A | 14 | 14 | Male | 115313 |
| ## 122 | Prof | A | 32 | 32 | Male | 124309 |
| ## 123 | Prof | A | 24 | 22 | Male | 97262 |
| ## 124 | AssocProf | A | 25 | 22 | Female | 62884 |
| ## 125 | Prof | A | 24 | 22 | Male | 96614 |
| ## 126 | Prof | A | 54 | 49 | Male | 78162 |
| ## 127 | Prof | A | 28 | 26 | Male | 155500 |
| ## 128 | AsstProf | A | 2 | 0 | Female | 72500 |
| ## 129 | Prof | A | 32 | 30 | Male | 113278 |
| ## 130 | AsstProf | A | 4 | 2 | Male | 73000 |
| ## 131 | AssocProf | A | 11 | 9 | Male | 83001 |
| ## 132 | Prof | A | 56 | 57 | Male | 76840 |
| ## 133 | AssocProf | A | 10 | 8 | Female | 77500 |

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|--------|-----------|---|----|----|--------|--------|
| ## 134 | AsstProf | A | 3 | 1 | Female | 72500 |
| ## 135 | Prof | A | 35 | 25 | Male | 168635 |
| ## 136 | Prof | A | 20 | 18 | Male | 136000 |
| ## 137 | Prof | A | 16 | 14 | Male | 108262 |
| ## 138 | Prof | A | 17 | 14 | Male | 105668 |
| ## 139 | AssocProf | A | 10 | 7 | Male | 73877 |
| ## 140 | Prof | A | 21 | 18 | Male | 152664 |
| ## 141 | AssocProf | A | 14 | 8 | Male | 100102 |
| ## 142 | AssocProf | A | 15 | 10 | Male | 81500 |
| ## 143 | Prof | A | 19 | 11 | Male | 106608 |
| ## 144 | AsstProf | B | 3 | 3 | Male | 89942 |
| ## 145 | Prof | B | 27 | 27 | Male | 112696 |
| ## 146 | Prof | B | 28 | 28 | Male | 119015 |
| ## 147 | AsstProf | B | 4 | 4 | Male | 92000 |
| ## 148 | Prof | B | 27 | 27 | Male | 156938 |
| ## 149 | Prof | B | 36 | 26 | Female | 144651 |
| ## 150 | AsstProf | B | 4 | 3 | Male | 95079 |
| ## 151 | Prof | B | 14 | 12 | Male | 128148 |
| ## 152 | AsstProf | B | 4 | 4 | Male | 92000 |
| ## 153 | Prof | B | 21 | 9 | Male | 111168 |
| ## 154 | AssocProf | B | 12 | 10 | Female | 103994 |
| ## 155 | AsstProf | B | 4 | 0 | Male | 92000 |
| ## 156 | Prof | B | 21 | 21 | Male | 118971 |
| ## 157 | AssocProf | B | 12 | 18 | Male | 113341 |
| ## 158 | AsstProf | B | 1 | 0 | Male | 88000 |
| ## 159 | AssocProf | B | 6 | 6 | Male | 95408 |
| ## 160 | Prof | B | 15 | 16 | Male | 137167 |
| ## 161 | AsstProf | B | 2 | 2 | Male | 89516 |
| ## 162 | Prof | B | 26 | 19 | Male | 176500 |
| ## 163 | AssocProf | B | 22 | 7 | Male | 98510 |
| ## 164 | AsstProf | B | 3 | 3 | Male | 89942 |
| ## 165 | AsstProf | B | 1 | 0 | Male | 88795 |
| ## 166 | Prof | B | 21 | 8 | Male | 105890 |
| ## 167 | Prof | B | 16 | 16 | Male | 167284 |
| ## 168 | Prof | B | 18 | 19 | Male | 130664 |
| ## 169 | AssocProf | B | 8 | 6 | Male | 101210 |
| ## 170 | Prof | B | 25 | 18 | Male | 181257 |
| ## 171 | AsstProf | B | 5 | 5 | Male | 91227 |
| ## 172 | Prof | B | 19 | 19 | Male | 151575 |
| ## 173 | Prof | B | 37 | 24 | Male | 93164 |
| ## 174 | Prof | B | 20 | 20 | Male | 134185 |
| ## 175 | AssocProf | B | 17 | 6 | Male | 105000 |
| ## 176 | Prof | B | 28 | 25 | Male | 111751 |
| ## 177 | AssocProf | B | 10 | 7 | Male | 95436 |
| ## 178 | AssocProf | B | 13 | 9 | Male | 100944 |
| ## 179 | Prof | B | 27 | 14 | Male | 147349 |
| ## 180 | AsstProf | B | 3 | 3 | Female | 92000 |
| ## 181 | Prof | B | 11 | 11 | Male | 142467 |
| ## 182 | Prof | B | 18 | 5 | Male | 141136 |
| ## 183 | AssocProf | B | 8 | 8 | Male | 100000 |

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|--------|-----------|---|----|----|--------|--------|
| ## 184 | Prof | B | 26 | 22 | Male | 150000 |
| ## 185 | Prof | B | 23 | 23 | Male | 101000 |
| ## 186 | Prof | B | 33 | 30 | Male | 134000 |
| ## 187 | AssocProf | B | 13 | 10 | Female | 103750 |
| ## 188 | Prof | B | 18 | 10 | Male | 107500 |
| ## 189 | AssocProf | B | 28 | 28 | Male | 106300 |
| ## 190 | Prof | B | 25 | 19 | Male | 153750 |
| ## 191 | Prof | B | 22 | 9 | Male | 180000 |
| ## 192 | Prof | B | 43 | 22 | Male | 133700 |
| ## 193 | Prof | B | 19 | 18 | Male | 122100 |
| ## 194 | AssocProf | B | 19 | 19 | Male | 86250 |
| ## 195 | AssocProf | B | 48 | 53 | Male | 90000 |
| ## 196 | AssocProf | B | 9 | 7 | Male | 113600 |
| ## 197 | AsstProf | B | 4 | 4 | Male | 92700 |
| ## 198 | AsstProf | B | 4 | 4 | Male | 92000 |
| ## 199 | Prof | B | 34 | 33 | Male | 189409 |
| ## 200 | Prof | B | 38 | 22 | Male | 114500 |
| ## 201 | AsstProf | B | 4 | 4 | Male | 92700 |
| ## 202 | Prof | B | 40 | 40 | Male | 119700 |
| ## 203 | Prof | B | 28 | 17 | Male | 160400 |
| ## 204 | Prof | B | 17 | 17 | Male | 152500 |
| ## 205 | Prof | B | 19 | 5 | Male | 165000 |
| ## 206 | Prof | B | 21 | 2 | Male | 96545 |
| ## 207 | Prof | B | 35 | 33 | Male | 162200 |
| ## 208 | Prof | B | 18 | 18 | Male | 120000 |
| ## 209 | AsstProf | B | 7 | 2 | Male | 91300 |
| ## 210 | Prof | B | 20 | 20 | Male | 163200 |
| ## 211 | AsstProf | B | 4 | 3 | Male | 91000 |
| ## 212 | Prof | B | 39 | 39 | Male | 111350 |
| ## 213 | Prof | B | 15 | 7 | Male | 128400 |
| ## 214 | Prof | B | 26 | 19 | Male | 126200 |
| ## 215 | AssocProf | B | 11 | 1 | Male | 118700 |
| ## 216 | Prof | B | 16 | 11 | Male | 145350 |
| ## 217 | Prof | B | 15 | 11 | Male | 146000 |
| ## 218 | AssocProf | B | 29 | 22 | Male | 105350 |
| ## 219 | AssocProf | B | 14 | 7 | Female | 109650 |
| ## 220 | Prof | B | 13 | 11 | Male | 119500 |
| ## 221 | Prof | B | 21 | 21 | Male | 170000 |
| ## 222 | Prof | B | 23 | 10 | Male | 145200 |
| ## 223 | AssocProf | B | 13 | 6 | Male | 107150 |
| ## 224 | Prof | B | 34 | 20 | Male | 129600 |
| ## 225 | Prof | A | 38 | 35 | Male | 87800 |
| ## 226 | Prof | A | 20 | 20 | Male | 122400 |
| ## 227 | AsstProf | A | 3 | 1 | Male | 63900 |
| ## 228 | AssocProf | A | 9 | 7 | Male | 70000 |
| ## 229 | Prof | A | 16 | 11 | Male | 88175 |
| ## 230 | Prof | A | 39 | 38 | Male | 133900 |
| ## 231 | Prof | A | 29 | 27 | Female | 91000 |
| ## 232 | AssocProf | A | 26 | 24 | Female | 73300 |
| ## 233 | Prof | A | 38 | 19 | Male | 148750 |

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|--------|-----------|---|----|----|--------|--------|
| ## 234 | Prof | A | 36 | 19 | Female | 117555 |
| ## 235 | AsstProf | A | 8 | 3 | Male | 69700 |
| ## 236 | Prof | A | 28 | 17 | Male | 81700 |
| ## 237 | Prof | A | 25 | 25 | Male | 114000 |
| ## 238 | AsstProf | A | 7 | 6 | Female | 63100 |
| ## 239 | Prof | A | 46 | 40 | Male | 77202 |
| ## 240 | Prof | A | 19 | 6 | Male | 96200 |
| ## 241 | AsstProf | A | 5 | 3 | Male | 69200 |
| ## 242 | Prof | A | 31 | 30 | Male | 122875 |
| ## 243 | Prof | A | 38 | 37 | Male | 102600 |
| ## 244 | Prof | A | 23 | 23 | Male | 108200 |
| ## 245 | Prof | A | 19 | 23 | Male | 84273 |
| ## 246 | Prof | A | 17 | 11 | Female | 90450 |
| ## 247 | Prof | A | 30 | 23 | Male | 91100 |
| ## 248 | Prof | A | 21 | 18 | Male | 101100 |
| ## 249 | Prof | A | 28 | 23 | Male | 128800 |
| ## 250 | Prof | A | 29 | 7 | Male | 204000 |
| ## 251 | Prof | A | 39 | 39 | Male | 109000 |
| ## 252 | Prof | A | 20 | 8 | Male | 102000 |
| ## 253 | Prof | A | 31 | 12 | Male | 132000 |
| ## 254 | AsstProf | A | 4 | 2 | Female | 77500 |
| ## 255 | Prof | A | 28 | 7 | Female | 116450 |
| ## 256 | AssocProf | A | 12 | 8 | Male | 83000 |
| ## 257 | Prof | A | 22 | 22 | Male | 140300 |
| ## 258 | AssocProf | A | 30 | 23 | Male | 74000 |
| ## 259 | AsstProf | A | 9 | 3 | Male | 73800 |
| ## 260 | Prof | A | 32 | 30 | Male | 92550 |
| ## 261 | AssocProf | A | 41 | 33 | Male | 88600 |
| ## 262 | Prof | A | 45 | 45 | Male | 107550 |
| ## 263 | Prof | A | 31 | 26 | Male | 121200 |
| ## 264 | Prof | A | 31 | 31 | Male | 126000 |
| ## 265 | Prof | A | 37 | 35 | Male | 99000 |
| ## 266 | Prof | A | 36 | 30 | Male | 134800 |
| ## 267 | Prof | A | 43 | 43 | Male | 143940 |
| ## 268 | Prof | A | 14 | 10 | Male | 104350 |
| ## 269 | Prof | A | 47 | 44 | Male | 89650 |
| ## 270 | Prof | A | 13 | 7 | Male | 103700 |
| ## 271 | Prof | A | 42 | 40 | Male | 143250 |
| ## 272 | Prof | A | 42 | 18 | Male | 194800 |
| ## 273 | AsstProf | A | 4 | 1 | Male | 73000 |
| ## 274 | AsstProf | A | 8 | 4 | Male | 74000 |
| ## 275 | AsstProf | A | 8 | 3 | Female | 78500 |
| ## 276 | Prof | A | 12 | 6 | Male | 93000 |
| ## 277 | Prof | A | 52 | 48 | Male | 107200 |
| ## 278 | Prof | A | 31 | 27 | Male | 163200 |
| ## 279 | Prof | A | 24 | 18 | Male | 107100 |
| ## 280 | Prof | A | 46 | 46 | Male | 100600 |
| ## 281 | Prof | A | 39 | 38 | Male | 136500 |
| ## 282 | Prof | A | 37 | 27 | Male | 103600 |
| ## 283 | Prof | A | 51 | 51 | Male | 57800 |

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|--------|-----------|---|----|----|--------|--------|
| ## 284 | Prof | A | 45 | 43 | Male | 155865 |
| ## 285 | AssocProf | A | 8 | 6 | Male | 88650 |
| ## 286 | AssocProf | A | 49 | 49 | Male | 81800 |
| ## 287 | Prof | A | 28 | 27 | Male | 115800 |
| ## 288 | AsstProf | A | 2 | 0 | Male | 85000 |
| ## 289 | Prof | A | 29 | 27 | Male | 150500 |
| ## 290 | AsstProf | A | 8 | 5 | Male | 74000 |
| ## 291 | Prof | A | 33 | 7 | Male | 174500 |
| ## 292 | Prof | A | 32 | 28 | Male | 168500 |
| ## 293 | Prof | A | 39 | 9 | Male | 183800 |
| ## 294 | AssocProf | A | 11 | 1 | Male | 104800 |
| ## 295 | Prof | A | 19 | 7 | Male | 107300 |
| ## 296 | Prof | A | 40 | 36 | Male | 97150 |
| ## 297 | Prof | A | 18 | 18 | Male | 126300 |
| ## 298 | Prof | A | 17 | 11 | Male | 148800 |
| ## 299 | Prof | A | 49 | 43 | Male | 72300 |
| ## 300 | AssocProf | A | 45 | 39 | Male | 70700 |
| ## 301 | Prof | A | 39 | 36 | Male | 88600 |
| ## 302 | Prof | A | 27 | 16 | Male | 127100 |
| ## 303 | Prof | A | 28 | 13 | Male | 170500 |
| ## 304 | Prof | A | 14 | 4 | Male | 105260 |
| ## 305 | Prof | A | 46 | 44 | Male | 144050 |
| ## 306 | Prof | A | 33 | 31 | Male | 111350 |
| ## 307 | AsstProf | A | 7 | 4 | Male | 74500 |
| ## 308 | Prof | A | 31 | 28 | Male | 122500 |
| ## 309 | AsstProf | A | 5 | 0 | Male | 74000 |
| ## 310 | Prof | A | 22 | 15 | Male | 166800 |
| ## 311 | Prof | A | 20 | 7 | Male | 92050 |
| ## 312 | Prof | A | 14 | 9 | Male | 108100 |
| ## 313 | Prof | A | 29 | 19 | Male | 94350 |
| ## 314 | Prof | A | 35 | 35 | Male | 100351 |
| ## 315 | Prof | A | 22 | 6 | Male | 146800 |
| ## 316 | AsstProf | B | 6 | 3 | Male | 84716 |
| ## 317 | AssocProf | B | 12 | 9 | Female | 71065 |
| ## 318 | Prof | B | 46 | 45 | Male | 67559 |
| ## 319 | Prof | B | 16 | 16 | Male | 134550 |
| ## 320 | Prof | B | 16 | 15 | Male | 135027 |
| ## 321 | Prof | B | 24 | 23 | Male | 104428 |
| ## 322 | AssocProf | B | 9 | 9 | Male | 95642 |
| ## 323 | AssocProf | B | 13 | 11 | Male | 126431 |
| ## 324 | Prof | B | 24 | 15 | Female | 161101 |
| ## 325 | Prof | B | 30 | 31 | Male | 162221 |
| ## 326 | AsstProf | B | 8 | 4 | Male | 84500 |
| ## 327 | Prof | B | 23 | 15 | Male | 124714 |
| ## 328 | Prof | B | 37 | 37 | Male | 151650 |
| ## 329 | AssocProf | B | 10 | 10 | Male | 99247 |
| ## 330 | Prof | B | 23 | 23 | Male | 134778 |
| ## 331 | Prof | B | 49 | 60 | Male | 192253 |
| ## 332 | Prof | B | 20 | 9 | Male | 116518 |
| ## 333 | Prof | B | 18 | 10 | Female | 105450 |

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|--------|-----------|---|----|----|--------|--------|
| ## 334 | Prof | B | 33 | 19 | Male | 145098 |
| ## 335 | AssocProf | B | 19 | 6 | Female | 104542 |
| ## 336 | Prof | B | 36 | 38 | Male | 151445 |
| ## 337 | Prof | B | 35 | 23 | Male | 98053 |
| ## 338 | Prof | B | 13 | 12 | Male | 145000 |
| ## 339 | Prof | B | 32 | 25 | Male | 128464 |
| ## 340 | Prof | B | 37 | 15 | Male | 137317 |
| ## 341 | Prof | B | 13 | 11 | Male | 106231 |
| ## 342 | Prof | B | 17 | 17 | Female | 124312 |
| ## 343 | Prof | B | 38 | 38 | Male | 114596 |
| ## 344 | Prof | B | 31 | 31 | Male | 162150 |
| ## 345 | Prof | B | 32 | 35 | Male | 150376 |
| ## 346 | Prof | B | 15 | 10 | Male | 107986 |
| ## 347 | Prof | B | 41 | 27 | Male | 142023 |
| ## 348 | Prof | B | 39 | 33 | Male | 128250 |
| ## 349 | AsstProf | B | 4 | 3 | Male | 80139 |
| ## 350 | Prof | B | 27 | 28 | Male | 144309 |
| ## 351 | Prof | B | 56 | 49 | Male | 186960 |
| ## 352 | Prof | B | 38 | 38 | Male | 93519 |
| ## 353 | Prof | B | 26 | 27 | Male | 142500 |
| ## 354 | Prof | B | 22 | 20 | Male | 138000 |
| ## 355 | AsstProf | B | 8 | 1 | Male | 83600 |
| ## 356 | Prof | B | 25 | 21 | Male | 145028 |
| ## 357 | Prof | A | 49 | 40 | Male | 88709 |
| ## 358 | Prof | A | 39 | 35 | Male | 107309 |
| ## 359 | Prof | A | 28 | 14 | Female | 109954 |
| ## 360 | AsstProf | A | 11 | 4 | Male | 78785 |
| ## 361 | Prof | A | 14 | 11 | Male | 121946 |
| ## 362 | Prof | A | 23 | 15 | Female | 109646 |
| ## 363 | Prof | A | 30 | 30 | Male | 138771 |
| ## 364 | AssocProf | A | 20 | 17 | Male | 81285 |
| ## 365 | Prof | A | 43 | 43 | Male | 205500 |
| ## 366 | Prof | A | 43 | 40 | Male | 101036 |
| ## 367 | Prof | A | 15 | 10 | Male | 115435 |
| ## 368 | AssocProf | A | 10 | 1 | Male | 108413 |
| ## 369 | Prof | A | 35 | 30 | Male | 131950 |
| ## 370 | Prof | A | 33 | 31 | Male | 134690 |
| ## 371 | AssocProf | A | 13 | 8 | Male | 78182 |
| ## 372 | Prof | A | 23 | 20 | Male | 110515 |
| ## 373 | Prof | A | 12 | 7 | Male | 109707 |
| ## 374 | Prof | A | 30 | 26 | Male | 136660 |
| ## 375 | Prof | A | 27 | 19 | Male | 103275 |
| ## 376 | Prof | A | 28 | 26 | Male | 103649 |
| ## 377 | AsstProf | A | 4 | 1 | Male | 74856 |
| ## 378 | AsstProf | A | 6 | 3 | Male | 77081 |
| ## 379 | Prof | A | 38 | 38 | Male | 150680 |
| ## 380 | AssocProf | A | 11 | 8 | Male | 104121 |
| ## 381 | AsstProf | A | 8 | 3 | Male | 75996 |
| ## 382 | Prof | A | 27 | 23 | Male | 172505 |
| ## 383 | AssocProf | A | 8 | 5 | Male | 86895 |

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|--------|----------|---|----|----|------|--------|
| ## 384 | Prof | A | 44 | 44 | Male | 105000 |
| ## 385 | Prof | A | 27 | 21 | Male | 125192 |
| ## 386 | Prof | A | 15 | 9 | Male | 114330 |
| ## 387 | Prof | A | 29 | 27 | Male | 139219 |
| ## 388 | Prof | A | 29 | 15 | Male | 109305 |
| ## 389 | Prof | A | 38 | 36 | Male | 119450 |
| ## 390 | Prof | A | 33 | 18 | Male | 186023 |
| ## 391 | Prof | A | 40 | 19 | Male | 166605 |
| ## 392 | Prof | A | 30 | 19 | Male | 151292 |
| ## 393 | Prof | A | 33 | 30 | Male | 103106 |
| ## 394 | Prof | A | 31 | 19 | Male | 150564 |
| ## 395 | Prof | A | 42 | 25 | Male | 101738 |
| ## 396 | Prof | A | 25 | 15 | Male | 95329 |
| ## 397 | AsstProf | A | 8 | 4 | Male | 81035 |

1.a.

```
fit_1a <- lm(salary ~ sex, data = Salaries)
fit_1a
```

```
##
## Call:
## lm(formula = salary ~ sex, data = Salaries)
##
## Coefficients:
## (Intercept)      sexMale
##      101002      14088
```

*# The general linear model with the response variable being 'salary' and a single predictor being 'sex' is: salary = 101002 + 14088 * I(sex = Male)*

```
summary(fit_1a)
```

```
##
## Call:
## lm(formula = salary ~ sex, data = Salaries)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -57290 -23502  -6828   19710 116455
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   101002      4809   21.001  < 2e-16 ***
## sexMale       14088       5065    2.782  0.00567 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 30030 on 395 degrees of freedom
## Multiple R-squared:  0.01921,    Adjusted R-squared:  0.01673
## F-statistic: 7.738 on 1 and 395 DF,  p-value: 0.005667
```

```
# p-value for sexMale is 0.005667, less than 0.05, reject H0.  
# Gender is a significant predictor of salary.
```

1.b.

```
Salaries$sex <- relevel(factor(Salaries$sex), ref = 'Male')  
fit_1b <- lm(salary ~ sex, data = Salaries)  
fit_1b  
  
##  
## Call:  
## lm(formula = salary ~ sex, data = Salaries)  
##  
## Coefficients:  
## (Intercept)      sexFemale  
##      115090      -14088  
  
# The general linear model with the response variable being 'salary' and a  
# single predictor being 'sex' with 'female' group as baseline is: salary =  
# 115090 - 14088 * I(sex = Female)
```

1.c.

```
fit_1c <- lm(salary ~ yrs.service + rank + discipline + sex, data = Salaries)  
anova(fit_1c)  
  
## Analysis of Variance Table  
##  
## Response: salary  
##           Df      Sum Sq   Mean Sq  F value    Pr(>F)        
## yrs.service  1 4.0709e+10 4.0709e+10   79.3405 < 2.2e-16 ***  
## rank         2 1.0358e+11 5.1789e+10  100.9335 < 2.2e-16 ***  
## discipline   1 1.7617e+10 1.7617e+10   34.3350 9.861e-09 ***  
## sex          1 7.7669e+08 7.7669e+08    1.5137  0.2193        
## Residuals   391 2.0062e+11 5.1310e+08        
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
  
# p-value for yrs.service is 0.426958, greater than the significance level,  
# 0.05, not reject H0. yrs.service is not significant.  
  
# p-value for rankAssocProf is 0.000428, less than the significance level,  
# 0.05, reject H0. rankAssocProf is significant.  
  
# p-value for rankProf is < 2e-16, less than the significance level, 0.05,  
# reject H0. rankProf is significant.  
  
# p-value for disciplineB is 1.24e-08, greater than the significance level,  
# 0.05, reject H0. disciplineB is not significant.  
  
# p-value for sexFemale is 0.219311, greater than the significance level,  
# 0.05, not reject H0. sexFemale is not significant.
```

```
summary(fit_1c)

##
## Call:
## lm(formula = salary ~ yrs.service + rank + discipline + sex,
##     data = Salaries)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -64202 -14255  -1533   10571   99163
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   73122.92    3245.27   22.532 < 2e-16 ***
## yrs.service    -88.78     111.64   -0.795  0.426958
## rankAssocProf 14560.40    4098.32    3.553  0.000428 ***
## rankProf      49159.64    3834.49   12.820 < 2e-16 ***
## disciplineB   13473.38    2315.50    5.819  1.24e-08 ***
## sexFemale     -4771.25    3878.00   -1.230  0.219311
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 22650 on 391 degrees of freedom
## Multiple R-squared:  0.4478, Adjusted R-squared:  0.4407
## F-statistic: 63.41 on 5 and 391 DF,  p-value: < 2.2e-16

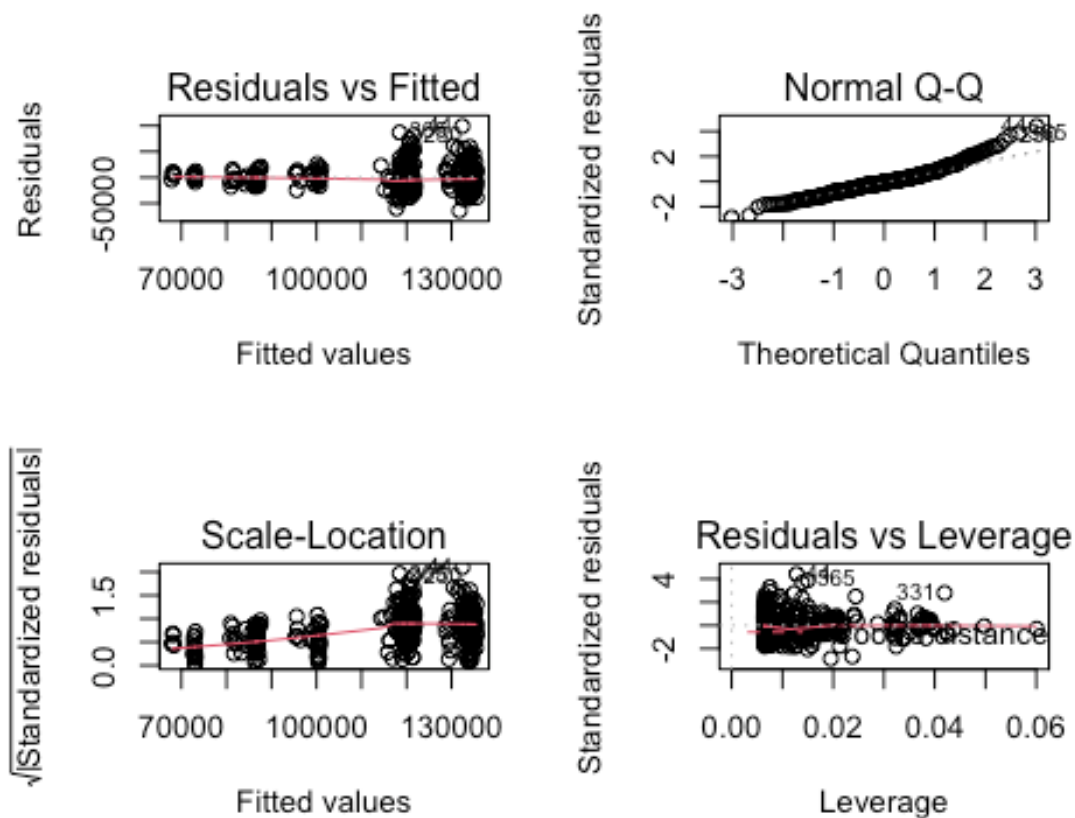
# LS of salary = 73122.92 - 88.78*yrs.service + 14560.40 * I(rank =
AssocProf) + 49159.64 * I(rank = Prof) + 13473.38 * I(discipline = B) -
4771.25 * I(sex = Female)
```

1.d.

```
# The coefficient of determination is 0.4478, which is greater than 0.2 and
less than 0.6
# This statistic indicates a good linear model fit
```

1.e.

```
par(mfrow = c(2,2))
plot(fit_1c)
```



1. Linearity: it is satisfied because the residuals are symmetrically distributed around the 0-line in the Residuals vs Fitted plot.

2. Homoscedasticity: it is not satisfied because the square root of standardized residuals is symmetrically distributed around the 1-line in the Scale-Location plot.

3. Independence: assume it is satisfied

4. Normality:

```
shapiro.test(residuals(fit_1c))
```

```
##
```

```
## Shapiro-Wilk normality test
```

```
##
```

```
## data: residuals(fit_1c)
```

```
## W = 0.96073, p-value = 8.202e-09
```

p-value is 8.202e-09, less than the significance level 0.05, so residuals are not normally distributed, normality assumption is not satisfied

Question 02

```
install.packages("tidyverse")
install.packages("caret")
install.packages("leaps")
install.packages("MASS")
library(tidyverse)

## — Attaching packages ————— tidyverse
1.3.1 —

## ✓ ggplot2 3.3.5      ✓ purrr  0.3.4
## ✓ tibble  3.1.4      ✓ dplyr  1.0.7
## ✓ tidyr   1.1.3      ✓ stringr 1.4.0
## ✓ readr   2.0.1      ✓ forcats 0.5.1

## — Conflicts —————
tidyverse_conflicts() —
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(caret)

## Loading required package: lattice

##
## Attaching package: 'caret'

## The following object is masked from 'package:purrr':
##
##   lift

library(leaps)
library(MASS)

##
## Attaching package: 'MASS'

## The following object is masked from 'package:dplyr':
##
##   select

data("swiss")
sample_n(swiss, 3)

##           Fertility Agriculture Examination Education Catholic
Infant.Mortality
## Le Locle      72.7           16.7           22           13      11.22
18.9
## Morges        65.5           59.8           22           10       5.23
18.0
```

| | | | | | |
|------------|------|------|----|---|------|
| ## Payerne | 74.2 | 58.1 | 14 | 8 | 5.23 |
| 23.8 | | | | | |

2.a.

```
models_2a <- regsubsets(Fertility ~ ., data = swiss, nvmax = 5)
summary(models_2a)
```

```
## Subset selection object
## Call: regsubsets.formula(Fertility ~ ., data = swiss, nvmax = 5)
## 5 Variables (and intercept)
##              Forced in Forced out
## Agriculture      FALSE      FALSE
## Examination      FALSE      FALSE
## Education         FALSE      FALSE
## Catholic          FALSE      FALSE
## Infant.Mortality  FALSE      FALSE
## 1 subsets of each size up to 5
## Selection Algorithm: exhaustive
##      Agriculture Examination Education Catholic Infant.Mortality
## 1  ( 1 ) " "          " "          "*"          " "          " "
## 2  ( 1 ) " "          " "          "*"          "*"          " "
## 3  ( 1 ) " "          " "          "*"          "*"          "*"
## 4  ( 1 ) "*"          " "          "*"          "*"          "*"
## 5  ( 1 ) "*"          "*"          "*"          "*"          "*"

# Best model with 1 variable: Fertility ~ Education
# Best model with 2 variables: Fertility ~ Education + Catholic
# Best model with 3 variables: Fertility ~ Education + Catholic +
Infant.Mortality
# Best model with 4 variables: Fertility ~ Agriculture + Education + Catholic
+ Infant.Mortality
# Best model with 5 variable: Fertility ~ Agriculture + Examination +
Education + Catholic + Infant.Mortality
```

2.b.

```
get_model_formula <- function(id, object, outcome){
  models <- summary(object)$which[id,-1]
  predictors <- names(which(models == TRUE))
  predictors <- paste(predictors, collapse = "+")
  as.formula(paste0(outcome, "~", predictors))
}

get_cv_error <- function(model.formula, data){
  set.seed(1)
  train.control <- trainControl(method = "cv", number = 5)
  cv <- train(model.formula, data = data, method = "lm",
              trControl = train.control)
  cv$results$RMSE
}
```

```

model.ids <- 1:5
cv.errors <- map(model.ids, get_model_formula, models_2a, "Fertility") %>%
  map(get_cv_error, data = swiss) %>%
  unlist()
cv.errors

## [1] 9.464156 8.517433 7.855267 7.601072 7.736328

which.min(cv.errors)

## [1] 4

coef(models_2a, 4)

##      (Intercept)      Agriculture      Education      Catholic
##      62.1013116      -0.1546175      -0.9802638       0.1246664
## Infant.Mortality
##      1.0784422

# The equation of the best overall model: Fertility = 62.1013116 -
# 0.1546175*Agriculture - 0.9802638*Education + 0.1246664*Cartholic +
# 1.0784422*Infant.Mortality

```

2.c.

```

stepAIC(lm(Fertility ~ Agriculture + Examination + Education + Catholic +
Infant.Mortality, data = swiss), direction = 'both', k = log(nrow(swiss)))

## Start:  AIC=201.79
## Fertility ~ Agriculture + Examination + Education + Catholic +
## Infant.Mortality
##
##           Df Sum of Sq  RSS   AIC
## - Examination    1    53.03 2158.1 199.11
## <none>                2105.0 201.79
## - Agriculture    1   307.72 2412.8 204.35
## - Infant.Mortality 1   408.75 2513.8 206.28
## - Catholic       1   447.71 2552.8 207.00
## - Education      1  1162.56 3267.6 218.61
##
## Step:  AIC=199.11
## Fertility ~ Agriculture + Education + Catholic + Infant.Mortality
##
##           Df Sum of Sq  RSS   AIC
## <none>                2158.1 199.11
## - Agriculture    1   264.18 2422.2 200.69
## + Examination    1    53.03 2105.0 201.79
## - Infant.Mortality 1   409.81 2567.9 203.43
## - Catholic       1   956.57 3114.6 212.50
## - Education      1  2249.97 4408.0 228.83

```

```
##
## Call:
## lm(formula = Fertility ~ Agriculture + Education + Catholic +
##      Infant.Mortality, data = swiss)
##
## Coefficients:
##      (Intercept)      Agriculture      Education      Catholic
##           62.1013          -0.1546          -0.9803           0.1247
## Infant.Mortality
##           1.0784

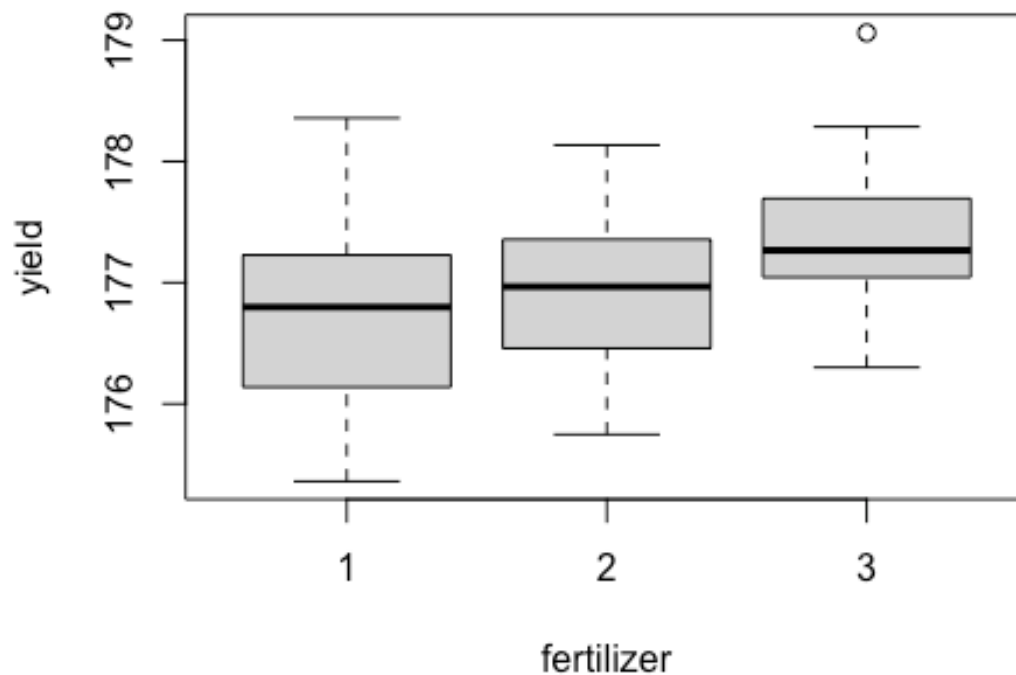
# The equation of the best overall model: Fertility = 62.1013 -
# 0.1546*Agriculture - 0.9803*Education + 0.1247*Catholic +
# 1.0784*Infant.Mortality
```

Question 03

```
crop_data <- read.csv('~\\OneDrive - Stony Brook University\\SBU\\MAT + AMS\\Fall
2021\\AMS 380\\hw\\04\\crop.data.csv', header = T)
```

3.a.

```
boxplot(yield ~ fertilizer, data = crop_data)
```



3.b.

```
fit_3b <- lm(yield ~ as.factor(fertilizer), data = crop_data)
anova(fit_3b)
```

Analysis of Variance Table

##

Response: yield

| | Df | Sum Sq | Mean Sq | F value | Pr(>F) |
|--------------------------|----|--------|---------|---------|---------------|
| ## as.factor(fertilizer) | 2 | 6.068 | 3.03402 | 7.8628 | 0.0006999 *** |
| ## Residuals | 93 | 35.886 | 0.38587 | | |

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

H0: $\mu_1 = \mu_2 = \mu_3$ Ha: at least one μ is different.

p-value for the F-test is 0.0006999, Less than the significance level 0.05, reject H0.

The effect of fertilizer is significant and the mean of different groups is different.

3.c.

```
res.aov <- aov(yield ~ factor(fertilizer), data = crop_data)
TukeyHSD(res.aov)
```

Tukey multiple comparisons of means

95% family-wise confidence level

##

Fit: aov(formula = yield ~ factor(fertilizer), data = crop_data)

##

\$`factor(fertilizer)`

| | diff | lwr | upr | p adj |
|--------|-----------|-------------|-----------|-----------|
| ## 2-1 | 0.1761687 | -0.19371896 | 0.5460564 | 0.4954705 |
| ## 3-1 | 0.5991256 | 0.22923789 | 0.9690133 | 0.0006125 |
| ## 3-2 | 0.4229569 | 0.05306916 | 0.7928445 | 0.0208735 |

p-value of comparison of fertilizer 1 and 2 is 0.4954705, greater than the significance level 0.05, not reject H0. The mean of fertilizer 1 and 2 are the same.

p-value of comparison of fertilizer 1 and 3 is 0.0006125, Less than the significance level 0.05, reject H0. The mean of fertilizer 1 and 3 are different.

p-value of comparison of fertilizer 2 and 3 is 0.0208735, Less than the significance level 0.05, reject H0. The mean of fertilizer 2 and 3 are different.

3.d.

H0: $\mu_2 = \mu_3$ Ha: $\mu_2 \neq \mu_3$

```

yield_2 <- crop_data$yield[crop_data$fertilizer == 2]
yield_3 <- crop_data$yield[crop_data$fertilizer == 3]

shapiro.test(yield_2)

##
##  Shapiro-Wilk normality test
##
## data:  yield_2
## W = 0.98329, p-value = 0.8875

shapiro.test(yield_3)

##
##  Shapiro-Wilk normality test
##
## data:  yield_3
## W = 0.95878, p-value = 0.2542

# p-value of the shapiro test of yield_2 and yield_3 are 0.8875 and 0.2542, both greater than the significance level 0.05, not reject H0, the samples are both normal.

var.test(yield_2,yield_3)

##
##  F test to compare two variances
##
## data:  yield_2 and yield_3
## F = 0.91811, num df = 31, denom df = 31, p-value = 0.8135
## alternative hypothesis: true ratio of variances is not equal to 1
## 95 percent confidence interval:
##  0.448169 1.880826
## sample estimates:
## ratio of variances
##          0.918111

# p-value of variance test is 0.8135, greater than the significance level 0.05, the variances of yield_2 and yield_3 are assumed to be the same.

t.test(yield_2, yield_3, mu = 0, var.equal = T)

##
##  Two Sample t-test
##
## data:  yield_2 and yield_3
## t = -2.8835, df = 62, p-value = 0.0054
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  -0.7161699 -0.1297438
## sample estimates:

```

```
## mean of x mean of y  
## 176.9332 177.3562
```

p-value of the t-test is 0.0054, less than the significance level 0.05, reject H_0 . The mean of yield of fertilizer 2 and 3 are significantly different.