Datenanalyse

Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Utilities Automotive Communications Financial Services Government Insurance Life Science & Healthcare Travel & Logistics Automotive Utilities

.consulting .solutions .partnership



Datenanalyse

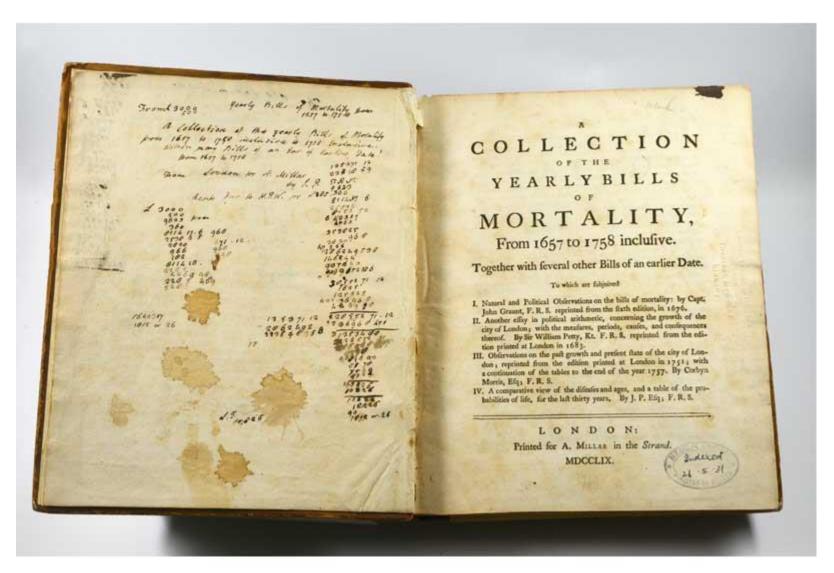


- Google´s Chef-Ökonom Hal Varian
 - "The next sexy job"
 - "The ability to take data to be able to understand it, to process it, to extract value from it, to communicate it that's going to be a hugely important skill."
 - New York Times, 2009
- "Hot new gig in tech" Fortune



Bills of Mortality





Bills of Mortality



| the Years of our Lord | rafa#li | 6451 | i fice | ident | 1551 | | | | | 100 | | | | A 1 | | | | 1632 | 1633 | 1634 | 161 | 11634 | 1630 | 1634 | 164 | 165 | 1655 1657 1658 | 619 | Year |
|--|---------------------|----------------|------------|------------|-----------------|---------------|-------------------|--------------------------|-------------------|-------------------|-------------------|----------------|----------------|-------------|-------------------|--------------|------------|--------------------|---------------------|----------------|--------|----------------|----------------------------------|----------------------|----------------------|----------------|------------------------------------|-------|-----------|
| er and Still-born | 335 | 129 | 317 889 | 351 898 | 389 780 | 381 834 | 384 864 282 | 433 974 1171 | 483 743 689 | 419 892 875 | 463 869 999 | 467 | 411 909 | 54+ 1095 | 499 579 856 | 439 712 | 410 661 | 445 671 1108 | 500 704 953 | 475 | 50 | 323 | 1793 2475 4418 | 2005 2814 5235 | 1341 3136 3865 | 1587 | 4812 1 3680 8 4363 4 445 | 377 | 237 |
| rs and Suddenly | 65 | 74 | 64 | 74 | 106 | 2 6 4 | 113 | 2 | 47 | 102 | 5 | 138 | 91 | 67 | 13 | 8 2 | 10 | 13 | 6 | 4 | 25 | + | 75 54 16 15 15 15 | 14 | 5 | 12 | 14 | 10 | 78 |
| Flux, Scouring and Flux and Scalded are a Geogreeic and Fiftula | 155 3 1 26 | 170 | 10 801 | 5 | 833 ## 51 | 763 | 5 1 36 | 38¢ 7 1 37 | 10 | 5 | 7 35 | + | 345 | 6 | 449 3 | 10 | 352 | 5 | 1 | 3 | 12 24 | 3 | ., | 4 | 24 | 4 | 20 3 150 1 | 19 | 6 |
| n Sore-mouth and Thrush led | 66 161 1369 | 1254 | 2000 | 117 | 1237 | 113 | 1050 | 1343 | 177 | 201 | 236 | 223 | 818 | 194 | 2595 | 157 | 112 | 171 2268 | 132 | 143 | 163 | 100 | 15 590 22778 105 | 451 | 190 498 678 | 769 | 161 1 859 4 4788 45 497 2 | 90 | 331 |
| Led Wind and Cough supplier and Cough school | 2422 | 1200 | 285 | 1088 | 1150 | LATO | 41 | 101 36 1865 318 | 2000 | 58 | 30 | 31 | 33 | 167 | 1817 | 58 | 51 | 55 | 45 | 54 | 50 | 57 | 174 | 107 | 9999 | 77 | 140 | 43 | 5 |
| p disk Stone c and Tympany and | | . 2 | 1 | 508 | 444 | 1 | \$17 \$17 | - | 4 | | 631 | 931 | 645 | 572 | 10 | | 179 | 280 34 | 0 1 166 37 | 5 250 31 | 315 | 359 45 | 5 C48 139 | 714 | 0 | | 972 130 215 13 | | 96 |
| ive drinking ned d in a Barb o-Sickneli | 1 | 17 | 1 | 45 | | 12 | 19 | 21 | 4 | 32 | - | | 4 | 18 | 19 | 13 | 7 | 18 | 13 | 20 % | 11 | 13 8 117 | 02 27 701 1 | 52 74 | 97 | 70 | 79 : | 2 2 | 31 |
| ed fmall Fox Ldcad in the Streets h-Fox | 15 | 400 6 29 | 1150 | . 4 | 325 | 1279 | 119 | 4 | 3 | 813 | 31 | 11 | 1523 | 354 | 18 | 33 | - 30 | - 6 | 13 | 17 | 121 | 14 22 3 | 85 51 2 | 247 | 19 80 | 34 81 5 | 116 | 9 3 5 | 1 |
| ed, and made-away themfolive | \$ 11 11 | | 12 16 | | 7 17 9 | 7 14 14 | 5 11 15 | 17 | 10 | 7 13 16 | 8 10 24 | 13 12 18 | 14 13 11 | | 18 | 5 10 8 | 32 6 | 11 | 14 | 5 17 3 | 17.000 | 7 2 | 71 17 0 | 50 | 48 48 48 | 25 59 47 | .63 | 5 : 6 | - 24 14 0 |

Wöchentliche Todesstatistiken



US Census Income Data

Explorative Datenanalyse

Daten



```
adult.data

1 39, State-gov, 77516, Bachelors, 13, Never-married, Adm-clerical, Not-in-family, White, Male, 2
50, Self-emp-not-inc, 83311, Bachelors, 13, Married-civ-spouse, Exec-managerial, Husband, White
3 38, Private, 215646, HS-grad, 9, Divorced, Handlers-cleaners, Not-in-family, White, Male, 0, 0,
4 53, Private, 234721, 11th, 7, Married-civ-spouse, Handlers-cleaners, Husband, Black, Male, 0, 0,
5 28, Private, 338409, Bachelors, 13, Married-civ-spouse, Prof-specialty, Wife, Black, Female, 0,
6 37, Private, 284582, Masters, 14, Married-civ-spouse, Exec-managerial, Wife, White, Female, 0,
7 49, Private, 160187, 9th, 5, Married-spouse-absent, Other-service, Not-in-family, Black, Female,
8 52, Self-emp-not-inc, 209642, HS-grad, 9, Married-civ-spouse, Exec-managerial, Husband, White,
9 31, Private, 45781, Masters, 14, Never-married, Prof-specialty, Not-in-family, White, Female, 1
10 42, Private, 159449, Bachelors, 13, Married-civ-spouse, Exec-managerial, Husband, White, Male,
11 37, Private, 280464, Some-college, 10, Married-civ-spouse, Exec-managerial, Husband, Black, Mal
12 30, State-gov, 141297, Bachelors, 13, Married-civ-spouse, Prof-specialty, Husband, Asian-Pac-Is
```

- Erste 12 Instanzen mit
- 15 Variablen

msg systems ag, 12. Juni 2013

Daten

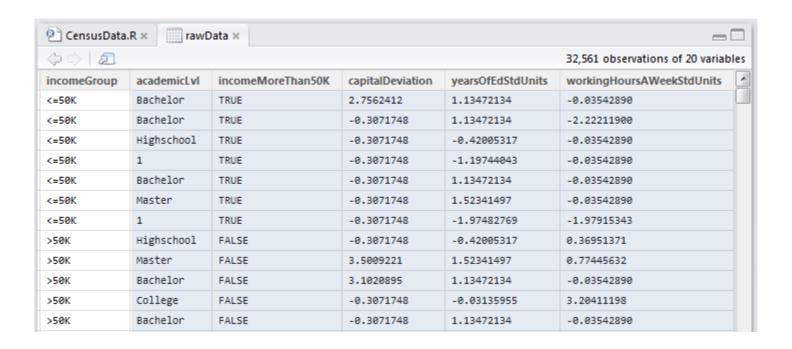


| 1 0 | 1 6 | | | | | | | |
|------------|-----|------------------|--------|--------------|-----------|-----------------------|-------------------|------------------------|
| þ þ | 2 | _ | | | | | 32,561 obse | ervations of 15 variab |
| | id | employerKind | fnlwgt | degree | yearsOfEd | maritalStatus | occupation | relationshipRole |
| 1 | 39 | State-gov | 77516 | Bachelors | 13 | Never-married | Adm-clerical | Not-in-family |
| 2 | 50 | Self-emp-not-inc | 83311 | Bachelors | 13 | Married-civ-spouse | Exec-managerial | Husband |
| 3 | 38 | Private | 215646 | HS-grad | 9 | Divorced | Handlers-cleaners | Not-in-family |
| 4 | 53 | Private | 234721 | 11th | 7 | Married-civ-spouse | Handlers-cleaners | Husband |
| 5 | 28 | Private | 338409 | Bachelors | 13 | Married-civ-spouse | Prof-specialty | Wife |
| 6 | 37 | Private | 284582 | Masters | 14 | Married-civ-spouse | Exec-managerial | Wife |
| 7 | 49 | Private | 160187 | 9th | 5 | Married-spouse-absent | Other-service | Not-in-family |
| 8 | 52 | Self-emp-not-inc | 209642 | HS-grad | 9 | Married-civ-spouse | Exec-managerial | Husband |
| 9 | 31 | Private | 45781 | Masters | 14 | Never-married | Prof-specialty | Not-in-family |
| 10 | 42 | Private | 159449 | Bachelors | 13 | Married-civ-spouse | Exec-managerial | Husband |
| 11 | 37 | Private | 280464 | Some-college | 10 | Married-civ-spouse | Exec-managerial | Husband |
| 12 | 30 | State-gov | 141297 | Bachelors | 13 | Married-civ-spouse | Prof-specialty | Husband |

- Erste 12 Instanzen als Data-Frame mit
- 8 von 15 Variablen

Daten



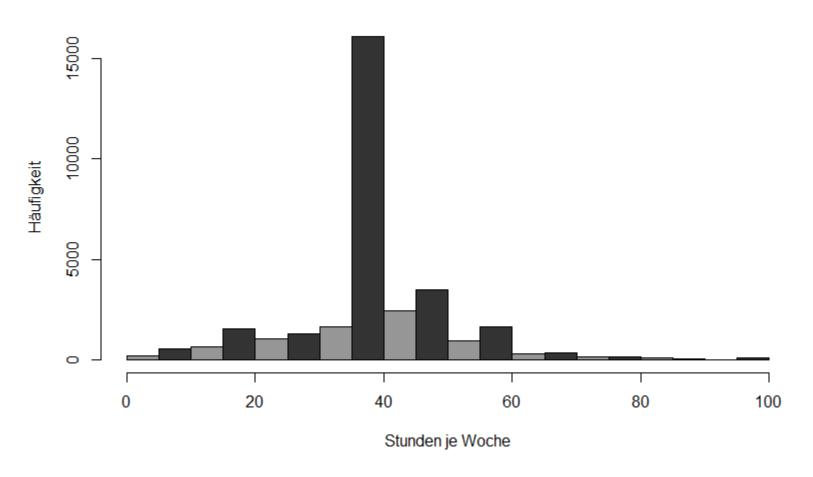


5 Sekundärvariablen

Visualisierungen



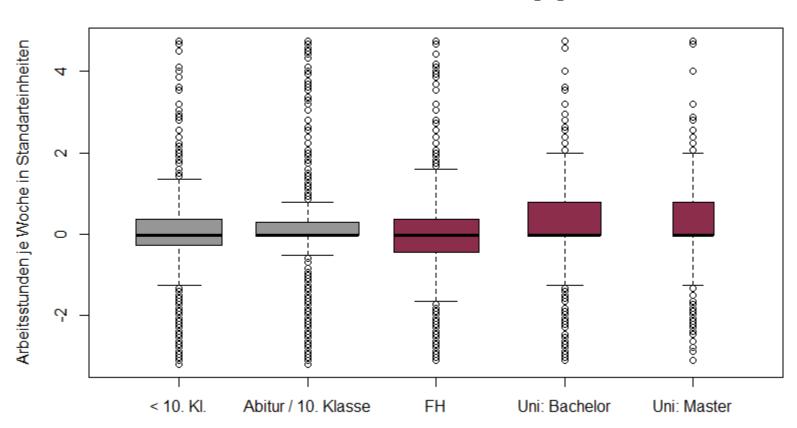




Visualisierungen



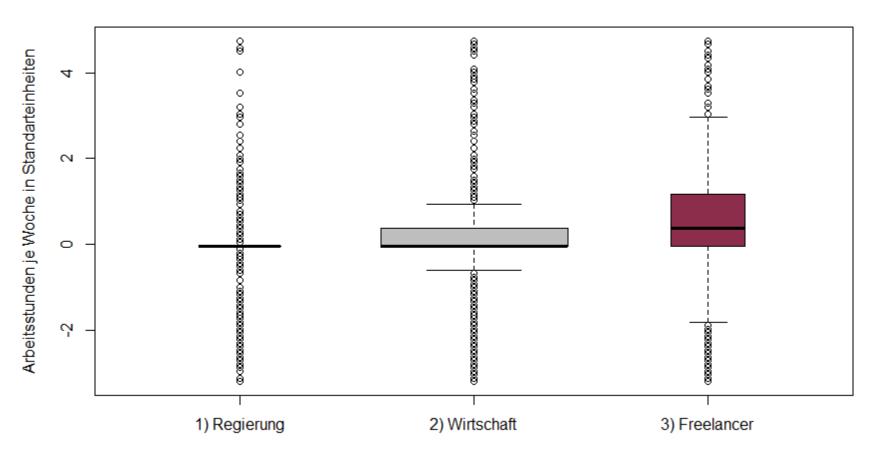
Arbeitsstunden nach Bildungsgrad



Visualisierungen

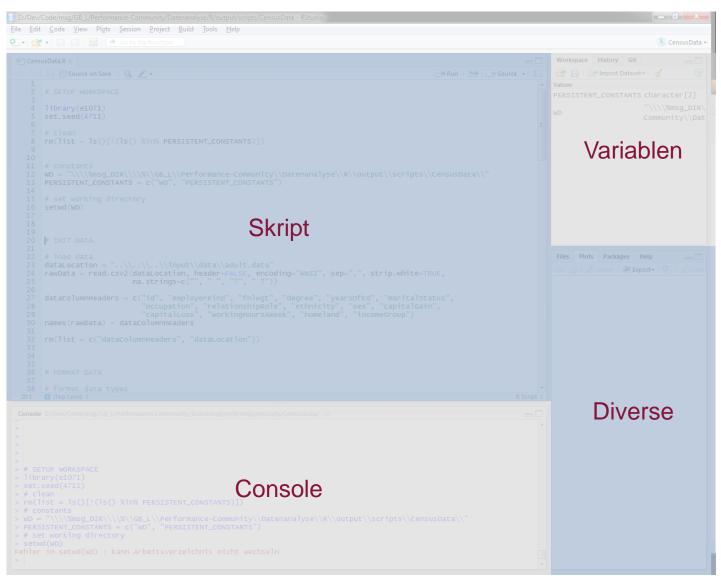


Arbeitsstunden nach Anstellungsart



RStudio





18

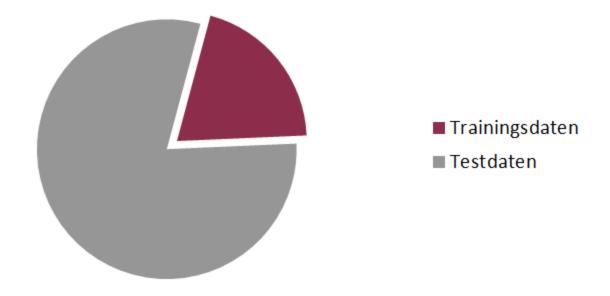


US Census Income Data

Klassifikation

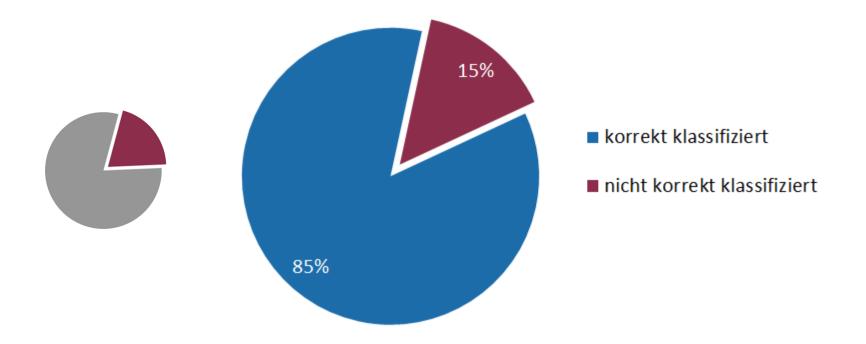
Klassifizierung





Klassifizierung





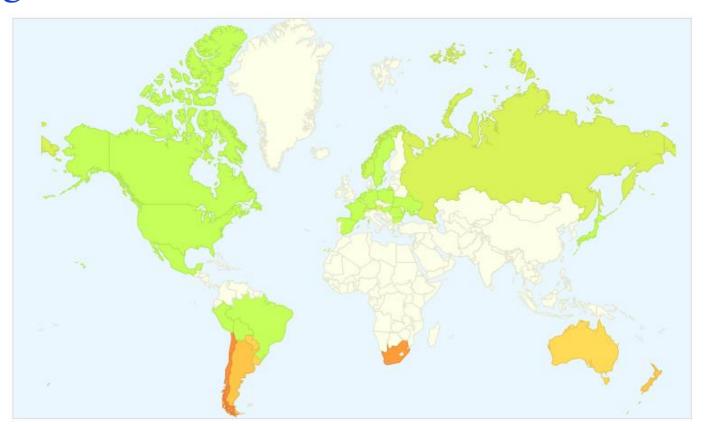


Anwendung

Anwendung



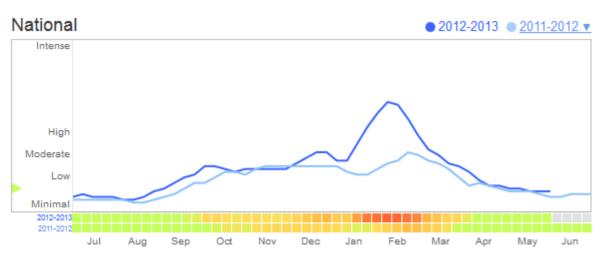
Google



Anwendung







Datenquellen





UCI Machine Learning Repository archive.ics.uci.edu/ml

Datenquellen



kaggle

kaggle.com

Ressourcen







Training Dataset for R Beginners

Last updated 3 minutes ago



AddHealth-Data-Analysis

The analysis of biases and influencer in attendance of religious services

Last updated a month ago

github.com/danielschulz/LearningR

Vielen Dank für Ihre Aufmerksamkeit



.consulting .solutions .partnership

