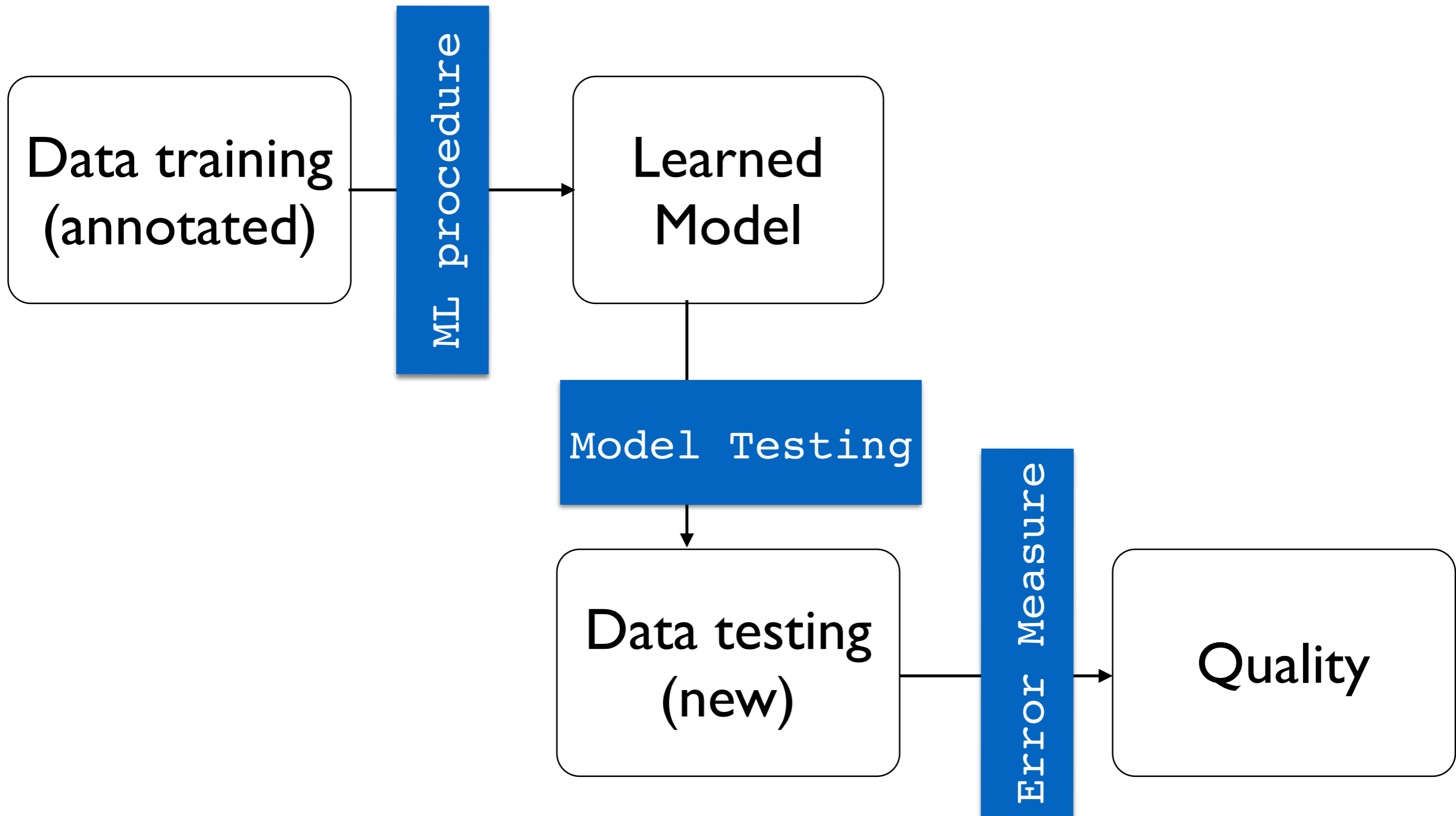


Intro to Machine Learning

Module 1 Objectives / Intro, Evaluation

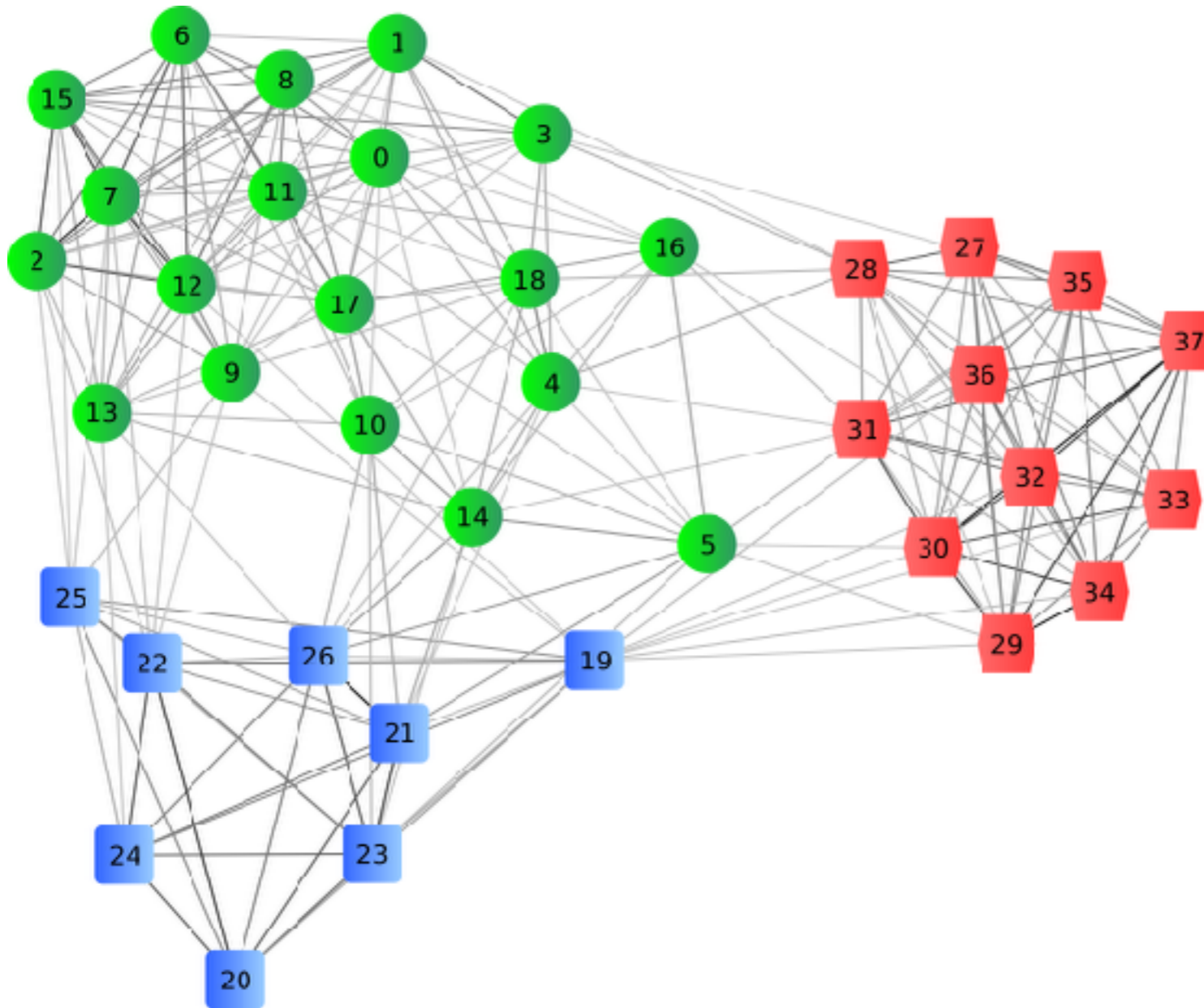
- Intro to Machine Learning - what is learning ?
- Data - Matrix type
 - algebraic notations
- Heuristics and Quantitative rules
- Error measurement
 - training VS testing error, Cross Validation
 - overfitting

What is machine learning ? Supervised learning



What is machine learning ? Graph learning

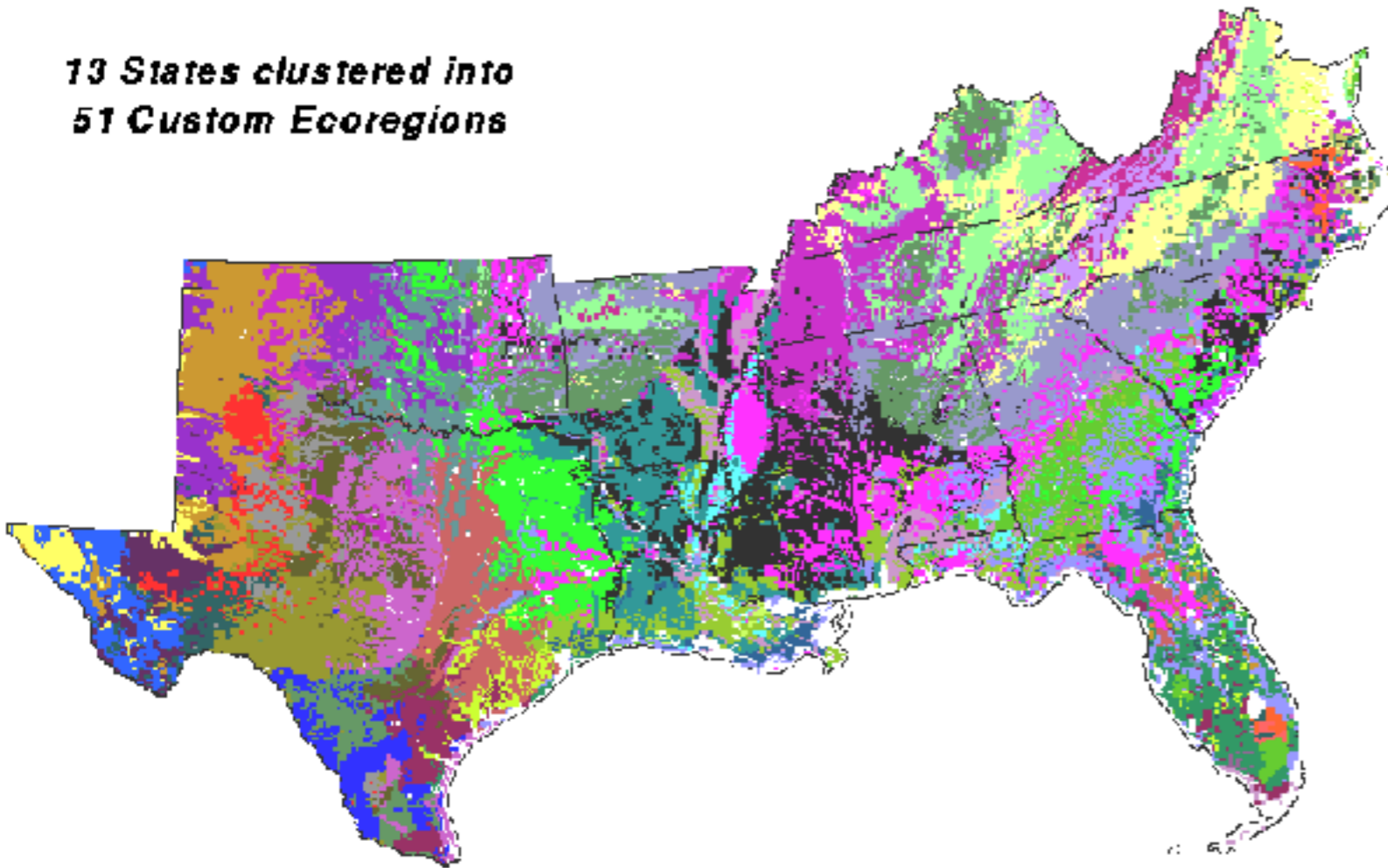
- data defined by links or analogies or connections
- for example social networks, or web links
- task: identify object properties from links
- tasks: detect graph patterns



What is machine learning ? Clustering

- data given without labels
- task: group similar data points

*13 States clustered into
51 Custom Ecoregions*

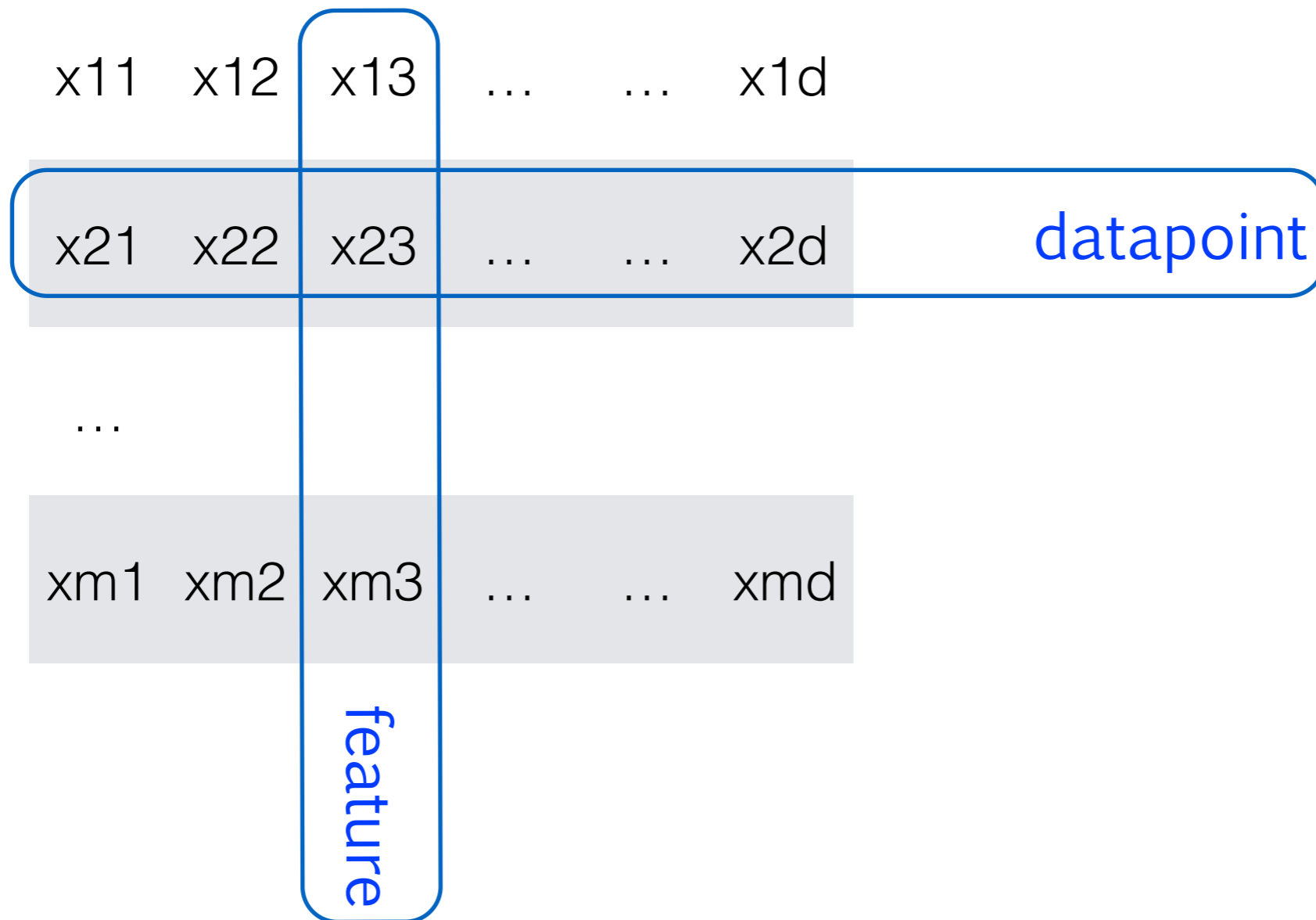


What is machine learning ? Time series analysis



- data that evolves with time
- like stocks or patient records
- task: predict future behavior
- task: detect anomalies

Matrix data



- m datapoints/objects $X=(x_1, x_2, \dots, x_d)$
- d features/columns f_1, f_2, \dots, f_d

Heuristic rules / decisional

- If fever > 100, patient has flu
- If email contains words “free” or “porn”, it is spam
- If a web page contains ngram “Michael Jackson”, it is relevant to the user
- If age < 22 and sex = F and highschool_diploma = Yes, then eligible for application
- If income_per_capita < \$1000, region prone to civil war
- If romantic = Yes and comedy = Yes and Orlando_Bloom = Yes, then movie success among females aged 20-40
- If Nasdaq_Computer_Index = Gain and Apple announces new Ipad, then AAPL_Stock = Buy

Heuristic rules / quantitative

- if $3 * \text{exam_grade} + 2 * \text{HW_grade} > 55$, then student can pass
- if $\text{blood_pressure} / \log(\text{age}) > 3$, recommend medicine
- if $\text{rent} + \text{food} + \text{bills} < 1/2 \text{ salary}$, loan for $1/2 \text{ salary}$ possible

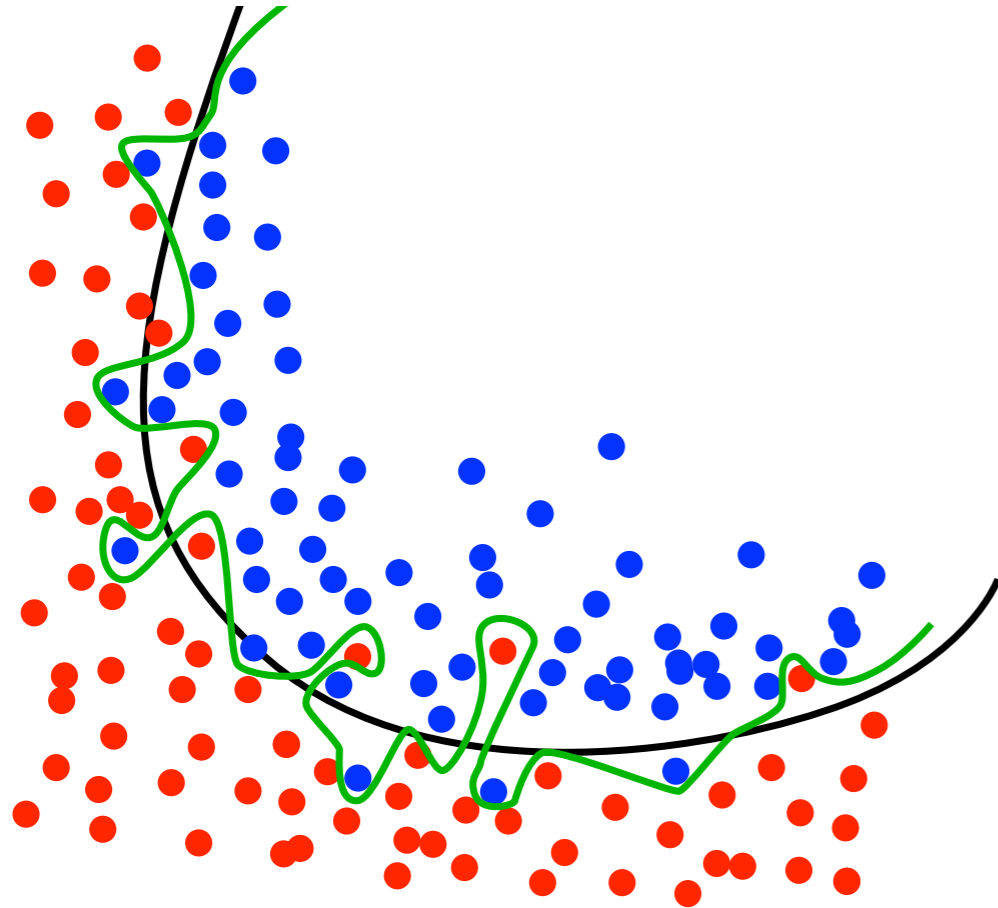
Matrix data / training VS testing

| | AUT | BEL | BUL | CYP | CZE | DEN | EST | FIN | FRA | GBR | GRE | HUN | IRL | ITA | LAT | LTU | LUX | MLT | NED | POL | POR | ROM | SVK | SLV | ESP | SWE | GBR |
|------|------|-------|-------|-------|-------|-------|-------|------|--------|--------|-------|-------|-------|--------|------|------|-------|-------|-------|-------|-------|-------|------|------|-------|-------|-------|
| T-01 | 64.4 | 185.0 | 44.7 | 7.0 | 124.1 | 51.3 | 14.9 | 66.6 | 363.5 | 837.4 | 92.2 | 65.8 | 42.8 | 446.6 | 6.6 | 11.6 | 2.4 | 2.1 | 174.8 | 303.8 | 64.8 | 30.7 | 35.9 | 15.1 | 304.9 | 49.8 | 579.2 |
| T-02 | 7.1 | 7.8 | 10.3 | 0.7 | 11.0 | 5.0 | 1.9 | 4.5 | 33.9 | 47.0 | 9.3 | 7.8 | 13.1 | 39.8 | 1.8 | 3.3 | 0.3 | 0.3 | 10.7 | 38.3 | 11.4 | 25.7 | 4.2 | 2.1 | 37.3 | 5.6 | 49.5 |
| T-03 | 5.3 | 11.0 | 4.4 | 0.7 | 8.0 | 7.0 | 0.8 | 6.9 | 72.3 | 66.3 | 9.1 | 9.7 | 8.8 | 40.5 | 1.5 | 5.0 | 0.4 | 0.3 | 17.6 | 31.1 | 8.1 | 18.8 | 3.7 | 1.3 | 28.8 | 7.7 | 39.6 |
| T-04 | 11.8 | 141 | 90 | 10 | 10 | 14 | 1.6 | 10 | 1,801 | 71.8 | 128 | 209 | 174 | 361 | 3 | 41 | 6 | 5 | 265 | 261 | 129 | 570 | 30 | 124 | 244 | 296 | 351 |
| T-05 | 91.2 | 1,454 | 387 | 91 | 594 | 806 | 8 | 854 | 10,958 | 9,353 | 1,162 | 51.6 | 431 | 5,267 | 19 | 19 | 63 | 42 | 1,354 | 2,750 | 361 | 4 | 175 | 95 | 5,011 | 777 | 9,221 |
| T-06 | 287 | 43 | 4 | 16 | 85 | 22 | 6 | 20 | 1,354 | 4,740 | 210 | 291 | 16 | 950 | 8 | 1 | 4 | 8 | 337 | 24 | 19 | 0 | 17 | 15 | 272 | 142 | 1,143 |
| T-07 | 644 | 1,230 | 447 | 70 | 1,241 | 513 | 149 | 596 | 3,635 | 8,374 | 922 | 598 | 428 | 4,406 | 65 | 116 | 84 | 21 | 1,748 | 3,068 | 648 | 907 | 978 | 151 | 3,049 | 488 | 5,682 |
| T-08 | 782 | 1,126 | 480 | 82 | 778 | 988 | 120 | 578 | 8,713 | 6,174 | 1,045 | 846 | 1,845 | 3,721 | 192 | 405 | 38 | 38 | 1,877 | 3,488 | 834 | 2,108 | 322 | 232 | 4,475 | 857 | 4,488 |
| T-09 | 228 | 135 | 848 | 28 | 291 | 137 | 63 | 244 | 1,410 | 1,388 | 328 | 294 | 178 | 1,933 | 78 | 154 | 3 | 12 | 864 | 1,221 | 647 | 740 | 211 | 88 | 1,298 | 218 | 2,308 |
| T-10 | 802 | 1,046 | 764 | 96 | 1,033 | 546 | 134 | 530 | 6,410 | 8,331 | 1,115 | 1,006 | 430 | 5,921 | 23 | 307 | 47 | 41 | 1,638 | 3,813 | 1,602 | 2,144 | 538 | 202 | 4,513 | 915 | 6,059 |
| T-11 | 305 | 11 | 112 | 8 | 125 | 108 | 89 | 297 | 418 | 1,188 | 43 | 83 | 16 | 338 | 87 | 59 | 1 | 1 | 95 | 732 | 47 | 56 | 110 | 15 | 496 | 318 | 253 |
| T-12 | 501 | 487 | 314 | 448 | 373 | 354 | 360 | 448 | 491 | 546 | 348 | 230 | 385 | 581 | 297 | 384 | 630 | 525 | 423 | 314 | 572 | 149 | 222 | 456 | 454 | 456 | 463 |
| T-13 | 282 | 641 | 131 | 53 | 308 | 171 | 60 | 220 | 1,970 | 2,650 | 436 | 138 | 182 | 1,881 | 47 | 55 | 62 | 19 | 347 | 445 | 332 | 212 | 74 | 63 | 1,573 | 362 | 1,827 |
| T-14 | 85.2 | 82.4 | 37.4 | 4.5 | 58.8 | 36.4 | 6.8 | 80.8 | 482.4 | 524.6 | 53.5 | 37.1 | 23.2 | 303.8 | 6.3 | 9.4 | 6.1 | 2.1 | 102.4 | 134.1 | 46.7 | 49.6 | 28.6 | 13.7 | 241.8 | 137.8 | 345.2 |
| T-15 | 9.00 | 17.06 | 3.47 | 0.01 | 9.80 | 4.82 | 1.44 | 4.88 | 48.41 | 102.00 | 2.34 | 14.48 | 4.30 | 80.81 | 1.91 | 2.92 | 1.38 | 0.00 | 81.30 | 18.87 | 4.30 | 18.00 | 6.00 | 1.10 | 27.00 | 0.98 | 98.47 |
| T-16 | 3.00 | 3.10 | 7.40 | 0.00 | 19.40 | 5.30 | 0.00 | 5.20 | 13.10 | 82.40 | 8.80 | 2.90 | 0.00 | 17.40 | 0.00 | 0.20 | 3.10 | 0.00 | 7.50 | 58.40 | 3.70 | 7.60 | 3.80 | 0.00 | 12.30 | 2.30 | 43.80 |
| T-17 | 389 | 989 | 98 | 88 | 388 | 385 | 8 | 77 | 10,978 | 3,433 | 889 | 770 | 233 | 16,880 | 53 | 60 | 55 | 30 | 1,492 | 950 | 1,265 | 270 | 250 | 890 | 3,402 | 178 | 3,313 |
| T-18 | 227 | 289 | 157 | 23 | 395 | 317 | 42 | 297 | 4,178 | 2,612 | 430 | 323 | 573 | 1,681 | 64 | 182 | 0 | 1 | 409 | 1,557 | 238 | 327 | 120 | 72 | 2,183 | 287 | 1,909 |
| T-19 | 3.5 | 5.8 | 2.3 | 3.2 | 3.8 | 3.6 | 3.3 | 6.4 | 6.4 | 6.1 | 2.5 | 2.4 | 3.9 | 3.0 | 1.5 | 2.0 | 2.4 | 2.1 | 4.8 | 3.3 | 2.5 | 1.6 | 3.2 | 3.3 | 3.1 | 5.4 | 4.0 |
| T-20 | 6.9 | 7.7 | 3.3 | 5.3 | 5.4 | 6.2 | 4.5 | 15.5 | 6.8 | 6.3 | 4.5 | 3.2 | 5.9 | 2.3 | 1.7 | 1.3 | 13.5 | 1.3 | 6.4 | 1.5 | 1.8 | 1.1 | 2.0 | 2.4 | 2.3 | 3.7 | 2.5 |
| T-21 | 0.46 | 3.43 | 0.19 | 0.03 | 6.43 | 1.01 | 0.08 | 0.19 | 0.99 | 1.82 | 0.47 | 0.23 | 0.45 | 1.03 | 0.04 | 0.21 | 0.03 | 0.00 | 1.51 | 0.28 | 0.22 | 0.27 | 0.27 | 0.17 | 0.44 | 0.27 | 2.76 |
| T-22 | 29 | 38 | 48 | 100 | 76 | 83 | 100 | 39 | 8 | 62 | 96 | 60 | 96 | 79 | 29 | 17 | 67 | 100 | 90 | 98 | 65 | 63 | 30 | 36 | 60 | 4 | 74 |
| T-23 | 133 | 178 | 7 | 13 | 41 | 111 | 8 | 138 | 780 | 782 | 105 | 32 | 164 | 295 | 11 | 10 | 38 | 15 | 227 | 72 | 96 | 26 | 2 | 13 | 518 | 234 | 963 |
| T-24 | 804 | 334 | 65 | 192 | 471 | 1,034 | 58 | 708 | 5,243 | 9,079 | 945 | 274 | 4,287 | 3,612 | 103 | 51 | 85 | 137 | 2,613 | 355 | 1,614 | 171 | 71 | 76 | 4,986 | 902 | 9,960 |
| T-25 | 130 | 103 | 7 | 0.00 | 53 | 73 | 7 | 87 | 860 | 1,070 | 80 | 46 | 197 | 398 | 7 | 10 | 74 | 22 | 429 | 68 | 138 | 36 | 5 | 13 | 473 | 128 | 977 |
| T-26 | 0.13 | 0.18 | 0.10 | 0.12 | 0.37 | 0.12 | 0.06 | 0.10 | 0.32 | 0.31 | 0.10 | 0.22 | 0.17 | 0.36 | 0.13 | 0.13 | 0.00 | 0.00 | 0.28 | 0.35 | 0.17 | 0.00 | 0.27 | 0.20 | 0.15 | 0.17 | 0.27 |
| T-27 | 639 | 464 | 493 | 739 | 280 | 737 | 436 | 468 | 543 | 691 | 438 | 450 | 740 | 542 | 310 | 373 | 736 | 511 | 624 | 245 | 446 | 382 | 259 | 423 | 507 | 482 | 584 |
| T-28 | 46 | 17 | 4 | 5 | 4 | 8 | 0 | 47 | 89 | 17 | 6 | 4 | 27 | 47 | 0 | 1 | 0 | 0 | 19 | 31 | 15 | 7 | 26 | 16 | 85 | 31 | 62 |
| T-29 | 521 | 628 | 1,004 | 3,711 | 1,379 | 843 | 1,234 | 897 | 162 | 1,140 | 2,247 | 976 | 2,421 | 1,473 | 362 | 139 | 1,707 | 2,858 | 1,575 | 1,501 | 1,377 | 744 | 298 | 851 | 1,248 | 41 | 1,170 |
| T-30 | 347 | 330 | 107 | 230 | 220 | 371 | 203 | 336 | 312 | 319 | 240 | 175 | 445 | 302 | 180 | 158 | 714 | 213 | 321 | 144 | 198 | 91 | 188 | 234 | 274 | 322 | 313 |
| T-31 | 0.0 | 0.0 | 20.2 | 4.8 | 0.6 | 7.0 | 0.6 | 0.2 | 20.5 | 0.1 | 18.3 | 1.3 | 0.5 | 75.4 | 1.0 | 0.1 | 0.1 | 0.1 | 0.2 | 49.5 | 111.8 | 1.6 | 0.1 | 0.7 | 92.4 | 1.3 | 0.2 |
| T-32 | 21.7 | 20.1 | 34.1 | 13.3 | 31.3 | 21.6 | 24.1 | 28.8 | 15.4 | 25.3 | 0.0 | 25.6 | 25.7 | 22.5 | 18.3 | 30.1 | 9.7 | 20.2 | 20.8 | 29.5 | 20.9 | 36.1 | 32.5 | 29.7 | 21.1 | 25.4 | 18.4 |
| T-33 | 134 | 117 | 34 | 8 | 127 | 72 | 13 | 51 | 951 | 231 | 107 | 70 | 16 | 480 | 23 | 60 | 5 | 2 | 105 | 249 | 59 | 98 | 57 | 20 | 657 | 167 | 372 |
| T-34 | 9.2 | 37.0 | 6.3 | 0.0 | 8.1 | 7.6 | 0.0 | 12.8 | 66.3 | 122.7 | 21.2 | 8.4 | 3.1 | 100.6 | 0.0 | 9.2 | 0.0 | 0.0 | 84.7 | 18.5 | 13.6 | 14.9 | 6.2 | 0.0 | 60.3 | 19.8 | 86.0 |
| T-35 | 1.0 | 3.2 | 0.5 | 0.1 | 1.4 | 0.3 | 7.3 | 2.5 | 7.6 | 20.0 | 0.3 | 1.4 | 0.7 | 6.1 | 0.0 | 0.1 | 0.1 | 0.0 | 1.9 | 1.6 | 2.3 | 2.2 | 0.4 | 0.1 | 3.1 | 1.2 | 8.0 |

Training
Testing

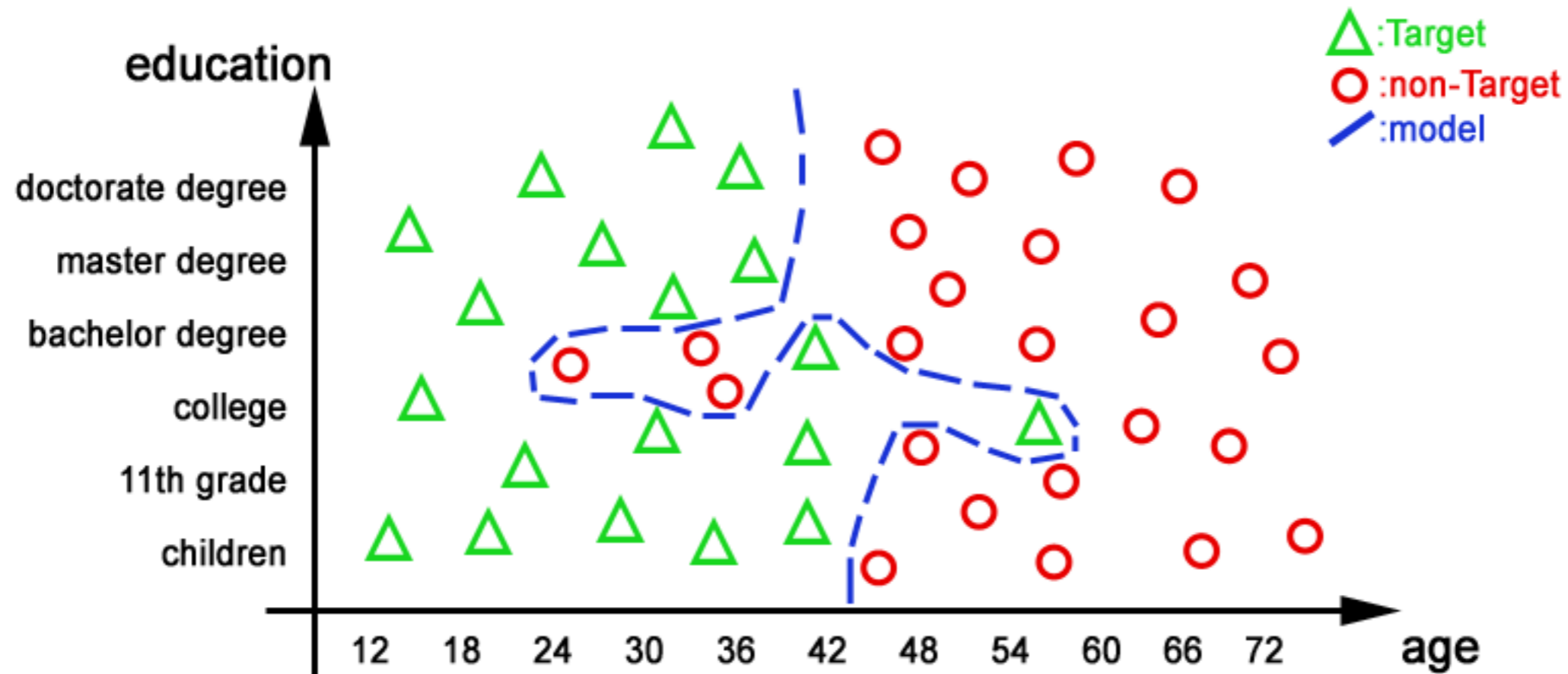
- testing set has to be independent of training set
 - or else testing result is inconclusive
 - and not reliable
- usually the data is partitioned before running any ML algorithm

Overfitting



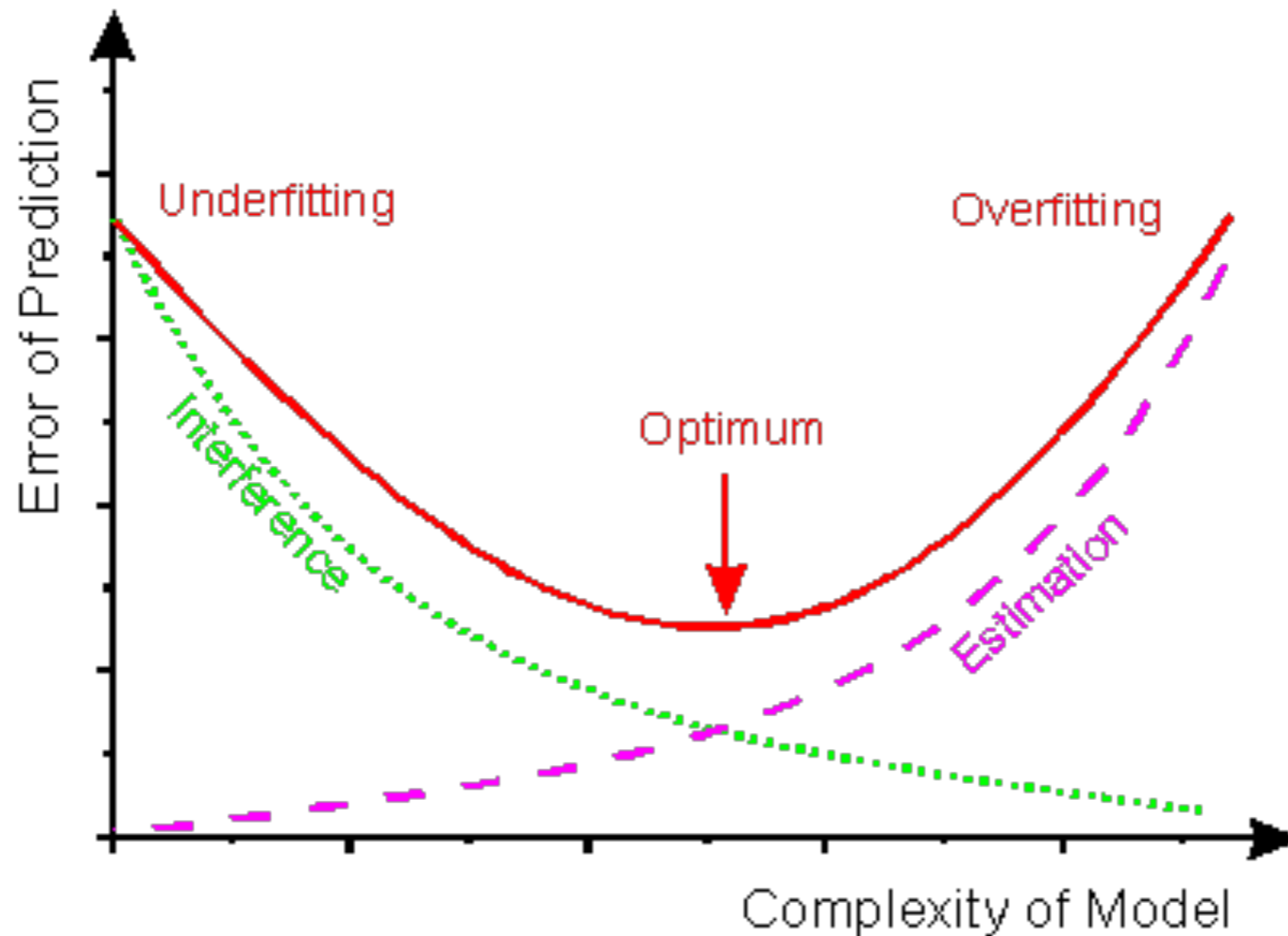
- might be capable to create a model that essentially memorizes all training dataset
 - for example a decision tree deep enough
- that is not useful : the purpose of the learning model is to applicable to new data (testing)

Overfitting



- might be capable to create a model that essentially memorizes all training dataset
 - for example a decision tree deep enough
- that is not useful : the purpose of the learning model is to applicable to new data (testing)

Overfitting



- as we keep training (insisting on ability to classify training set), the performance on the training set (green) becomes unrealistically small
 - model becomes more complex
- but at the same time ability to predict/classify new data (pink) worsens

Cross Validation Setup

| | AUT | BEL | BUL | CYP | CZE | DEN | EST | FIN | FRA | GER | GRE | HUN | IRL | ITA | LAT | LTU | LUX | MUT | NED | POL | POR | ROM | SVK | SLO | ESP | SWE | GBR |
|------|------|------|------|------|-------|------|------|------|-------|-------|------|------|------|-------|------|------|------|------|------|-------|-------|------|------|------|-------|-------|-------|
| T-01 | 644 | 1250 | 467 | 70 | 1241 | 513 | 149 | 586 | 3835 | 8374 | 322 | 538 | 423 | 4455 | 65 | 118 | 84 | 21 | 1748 | 3038 | 848 | 907 | 369 | 151 | 3049 | 458 | 538.2 |
| T-02 | 7.1 | 7.8 | 10.3 | 0.7 | 11.0 | 5.5 | 1.9 | 4.5 | 56.9 | 47.6 | 9.3 | 7.9 | 13.1 | 39.8 | 1.8 | 3.3 | 0.3 | 0.3 | 16.7 | 38.3 | 11.4 | 25.7 | 4.2 | 2.1 | 37.3 | 5.6 | 48.5 |
| T-03 | 5.3 | 11.0 | 4.4 | 0.7 | 8.0 | 7.0 | 0.8 | 6.9 | 72.3 | 63.5 | 9.1 | 9.7 | 8.8 | 40.5 | 1.5 | 5.0 | 0.4 | 0.3 | 17.6 | 31.1 | 6.1 | 16.8 | 3.7 | 1.3 | 29.8 | 7.7 | 38.6 |
| T-04 | 119 | 141 | 90 | 10 | 10 | 14 | 16 | 10 | 1301 | 718 | 128 | 208 | 174 | 361 | 3 | 41 | 6 | 5 | 285 | 281 | 129 | 570 | 20 | 124 | 244 | 285 | 351 |
| T-05 | 912 | 1464 | 387 | 91 | 894 | 805 | 8 | 864 | 10938 | 9383 | 1162 | 518 | 431 | 5287 | 19 | 19 | 82 | 42 | 1354 | 2790 | 324 | 4 | 175 | 95 | 5011 | 777 | 9221 |
| T-06 | 267 | 43 | 4 | 16 | 86 | 22 | 6 | 20 | 1354 | 4740 | 210 | 261 | 56 | 481 | 8 | 1 | 4 | 8 | 337 | 24 | 10 | 0 | 17 | 19 | 272 | 142 | 1143 |
| T-07 | 844 | 1280 | 447 | 70 | 1241 | 513 | 149 | 586 | 3835 | 8374 | 322 | 538 | 423 | 4455 | 65 | 118 | 84 | 21 | 1748 | 3038 | 848 | 907 | 369 | 151 | 3049 | 458 | 538.2 |
| T-08 | 782 | 1126 | 485 | 82 | 779 | 583 | 120 | 758 | 9330 | 6354 | 1045 | 846 | 1845 | 3721 | 192 | 405 | 38 | 38 | 1817 | 3488 | 824 | 2028 | 322 | 202 | 4476 | 857 | 4458 |
| T-09 | 228 | 133 | 648 | 25 | 291 | 137 | 83 | 244 | 1410 | 1389 | 328 | 284 | 178 | 1593 | 75 | 184 | 3 | 12 | 684 | 1221 | 847 | 740 | 211 | 65 | 1298 | 215 | 2208 |
| T-10 | 832 | 1046 | 784 | 85 | 1020 | 545 | 134 | 530 | 6410 | 8231 | 1115 | 1005 | 430 | 3521 | 23 | 337 | 47 | 41 | 1038 | 3815 | 1082 | 2144 | 518 | 202 | 4512 | 915 | 6058 |
| T-11 | 305 | 11 | 112 | 8 | 125 | 109 | 89 | 287 | 619 | 1165 | 48 | 83 | 16 | 338 | 97 | 58 | 4 | 1 | 95 | 732 | 47 | 68 | 110 | 15 | 456 | 319 | 266 |
| T-12 | 501 | 467 | 314 | 448 | 373 | 354 | 350 | 448 | 891 | 546 | 348 | 281 | 385 | 581 | 287 | 384 | 679 | 325 | 428 | 314 | 572 | 148 | 222 | 458 | 454 | 458 | 463 |
| T-13 | 282 | 641 | 131 | 63 | 938 | 171 | 81 | 220 | 1970 | 2580 | 495 | 138 | 188 | 1881 | 47 | 56 | 68 | 19 | 547 | 446 | 382 | 812 | 74 | 63 | 1573 | 382 | 1822 |
| T-14 | 652 | 824 | 374 | 45 | 568 | 364 | 68 | 818 | 4824 | 5345 | 315 | 371 | 232 | 3038 | 63 | 84 | 61 | 21 | 1024 | 1241 | 451 | 485 | 255 | 137 | 281.8 | 137.8 | 345.2 |
| T-15 | 910 | 1706 | 347 | 0.01 | 960 | 482 | 144 | 436 | 4541 | 10240 | 234 | 1445 | 430 | 8051 | 1.91 | 298 | 136 | 0.00 | 5130 | 1567 | 430 | 1300 | 600 | 1.10 | 2701 | 0.98 | 9847 |
| T-16 | 3.00 | 3.10 | 7.40 | 0.00 | 13.40 | 5.50 | 0.00 | 5.20 | 13.10 | 82.40 | 8.80 | 2.90 | 0.00 | 17.40 | 0.00 | 0.20 | 3.10 | 0.00 | 7.50 | 58.40 | 3.70 | 7.80 | 3.80 | 0.00 | 18.30 | 2.20 | 43.80 |
| T-17 | 360 | 989 | 98 | 89 | 388 | 385 | 8 | 77 | 10979 | 3463 | 990 | 770 | 238 | 16360 | 53 | 60 | 36 | 30 | 1432 | 950 | 1246 | 270 | 280 | 550 | 3402 | 179 | 3313 |
| T-18 | 227 | 280 | 157 | 23 | 395 | 317 | 42 | 287 | 4178 | 2512 | 420 | 323 | 573 | 1481 | 64 | 162 | 0 | 1 | 408 | 1557 | 228 | 327 | 120 | 72 | 2183 | 287 | 1800 |
| T-19 | 3.5 | 5.8 | 2.3 | 3.2 | 3.9 | 3.6 | 3.3 | 6.4 | 6.4 | 6.1 | 2.6 | 2.4 | 3.9 | 3.0 | 1.5 | 2.0 | 8.4 | 2.1 | 4.8 | 2.3 | 2.5 | 1.6 | 3.2 | 3.3 | 3.1 | 5.4 | 4.0 |
| T-20 | 6.9 | 7.7 | 3.3 | 5.3 | 5.4 | 6.2 | 4.5 | 15.5 | 5.8 | 6.3 | 4.5 | 3.2 | 5.9 | 2.3 | 1.7 | 1.3 | 13.5 | 1.3 | 6.4 | 1.5 | 1.8 | 1.1 | 2.0 | 2.4 | 2.3 | 3.7 | 2.5 |
| T-21 | 0.16 | 3.83 | 0.18 | 0.00 | 0.83 | 1.01 | 0.08 | 0.18 | 0.88 | 1.82 | 0.17 | 0.23 | 0.45 | 1.00 | 0.04 | 0.21 | 0.00 | 0.00 | 1.51 | 0.28 | 0.23 | 0.27 | 0.27 | 0.17 | 0.14 | 0.27 | 3.76 |
| T-22 | 29 | 38 | 48 | 140 | 78 | 85 | 160 | 39 | 8 | 82 | 95 | 89 | 98 | 79 | 29 | 17 | 57 | 109 | 90 | 46 | 55 | 63 | 50 | 35 | 50 | 4 | 74 |
| T-23 | 133 | 178 | 7 | 13 | 41 | 111 | 8 | 128 | 786 | 782 | 103 | 32 | 164 | 386 | 11 | 10 | 38 | 15 | 227 | 72 | 96 | 20 | 2 | 13 | 518 | 234 | 985 |
| T-24 | 804 | 334 | 65 | 192 | 471 | 1034 | 58 | 706 | 5248 | 9079 | 945 | 274 | 4287 | 3612 | 100 | 51 | 85 | 127 | 2513 | 355 | 1014 | 171 | 71 | 75 | 4586 | 902 | 9360 |
| T-25 | 130 | 106 | 7 | 0.00 | 59 | 78 | 7 | 87 | 860 | 1070 | 80 | 66 | 197 | 388 | 7 | 10 | 74 | 22 | 428 | 68 | 128 | 26 | 5 | 13 | 073 | 128 | 877 |
| T-26 | 0.13 | 0.19 | 0.10 | 0.12 | 0.57 | 0.12 | 0.05 | 0.10 | 0.32 | 0.31 | 0.10 | 0.22 | 0.17 | 0.38 | 0.13 | 0.14 | 0.19 | 0.10 | 0.28 | 0.35 | 0.17 | 0.10 | 0.27 | 0.29 | 0.15 | 0.17 | 0.27 |
| T-27 | 630 | 464 | 483 | 738 | 288 | 737 | 431 | 468 | 543 | 601 | 438 | 458 | 740 | 542 | 310 | 378 | 705 | 611 | 684 | 245 | 445 | 382 | 288 | 423 | 597 | 482 | 584 |
| T-28 | 46 | 17 | 4 | 5 | 4 | 8 | 0 | 47 | 59 | 17 | 8 | 4 | 27 | 47 | 0 | 1 | 0 | 0 | 19 | 31 | 15 | 7 | 28 | 18 | 85 | 31 | 62 |
| T-29 | 521 | 628 | 1094 | 3711 | 1358 | 843 | 1254 | 1897 | 162 | 1140 | 2247 | 876 | 2423 | 1473 | 382 | 139 | 1707 | 2386 | 1575 | 1501 | 1377 | 744 | 788 | 851 | 1288 | 41 | 1170 |
| T-30 | 347 | 330 | 107 | 230 | 220 | 371 | 308 | 335 | 312 | 319 | 240 | 175 | 445 | 302 | 160 | 183 | 714 | 213 | 321 | 144 | 198 | 91 | 188 | 234 | 274 | 322 | 318 |
| T-31 | 0.0 | 0.0 | 20.2 | 4.8 | 0.6 | 7.0 | 0.6 | 0.2 | 20.5 | 0.1 | 18.3 | 1.3 | 0.5 | 76.4 | 1.0 | 0.1 | 0.1 | 0.1 | 0.2 | 49.5 | 111.8 | 1.5 | 0.1 | 0.7 | 92.4 | 1.3 | 0.2 |
| T-32 | 247 | 26.1 | 34.1 | 13.3 | 34.3 | 21.8 | 24.1 | 28.8 | 18.4 | 28.3 | 0.0 | 29.6 | 26.7 | 22.8 | 18.3 | 30.1 | 9.7 | 29.2 | 20.8 | 29.5 | 20.9 | 38.1 | 32.5 | 29.7 | 21.1 | 28.4 | 18.4 |
| T-33 | 124 | 117 | 34 | 8 | 127 | 72 | 13 | 51 | 881 | 231 | 107 | 79 | 56 | 481 | 28 | 28 | 3 | 2 | 36 | 248 | 39 | 58 | 27 | 20 | 607 | 107 | 272 |
| T-34 | 92 | 37.0 | 6.3 | 0.0 | 8.1 | 7.5 | 0.0 | 12.8 | 88.3 | 122.7 | 21.2 | 8.4 | 3.1 | 103.6 | 0.0 | 9.2 | 0.0 | 0.0 | 84.7 | 18.5 | 13.6 | 14.9 | 82 | 0.0 | 60.3 | 19.8 | 80.0 |
| T-35 | 1.0 | 5.2 | 0.5 | 0.1 | 1.4 | 0.3 | 7.3 | 2.5 | 7.6 | 23.0 | 0.3 | 1.4 | 0.7 | 6.1 | 0.0 | 0.1 | 0.1 | 0.0 | 1.9 | 1.6 | 2.3 | 2.2 | 0.4 | 0.1 | 3.1 | 1.2 | 8.0 |

Fold 1

Fold 2

Fold 3

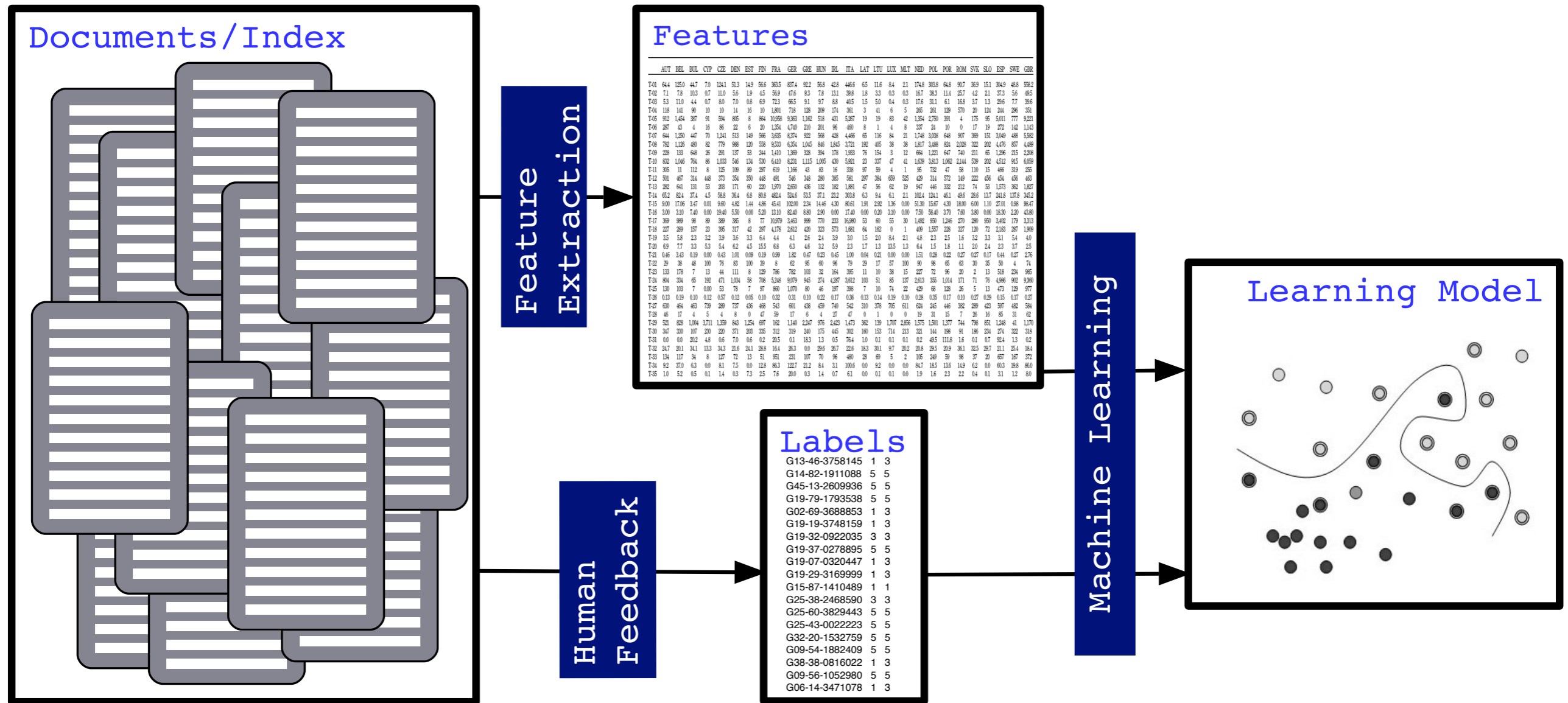
Fold 4

Fold K-1

Fold K

- split data in K folds
- execute K independent learning trials:
 - train on K-1 folds
 - test on remaining fold
 - measure testing performance
- average results across K trials

Learning / Training with text objects



- for objects like text documents or images:
 - extract features (to obtain matrix form)
 - annotate (to obtain labels)

Spambase dataset

- about 4000 emails
- 54 features numerical
- two classes: spam / no_spam

Housing dataset

- 1300 houses
- 13 features (numerical)
- label : purchase prices (quantitative)

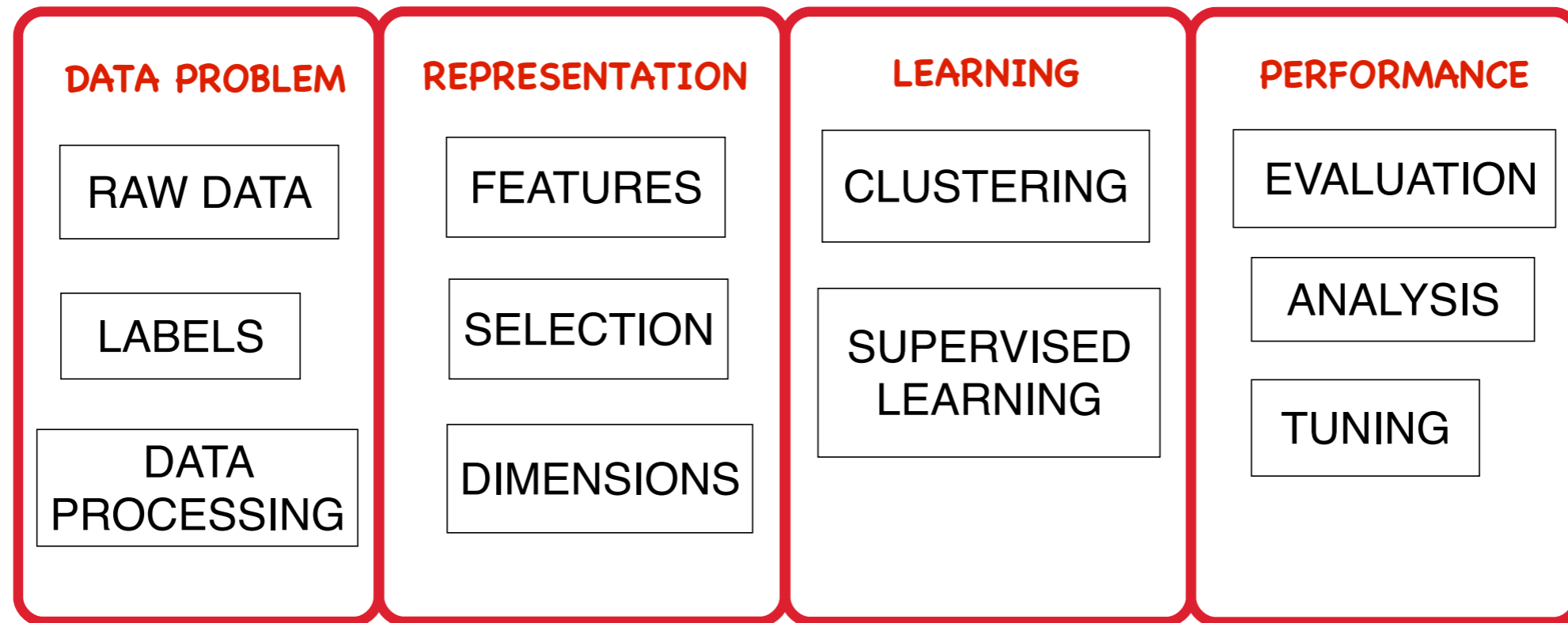
Digits dataset

- 60000 images of scanned digits
- 28x28 pixel per image, black or white
- features not extracted
- 10 classes : 0,1,2, ..., 9

Documents dataset

- 20,000 news articles (text)
- features not extracted
- 20 categories: religion, music, computers, sports, etc.

course map



- main focus: learning algorithms
- main focus: hands-on practice on datasets
- secondary focus: analysis, error measurement
- secondary focus: features, representation

typical module subtasks / objectives

- THEORY
 - explain/understand fundamental mechanism
 - proof (math, intuition)
 - pseudocode
- CODE
 - run existing code
 - implement and demo your code
 - data handling: features, dimensionality, scale, missing values, normalization
 - computational issues : memory, cache, CPU, disk
- EVALUATION
 - setup
 - performance measurement, comparison
 - analysis/failure of procedure behavior
- HOWTO
 - practical advise, hacks, heuristics
 - communicate on topic well : email, forums
 - where to look online