

# Untitled

2024-11-17

```
# install.packages('catdata')
```

```
library(catdata)
```

```
## Loading required package: MASS
```

```
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following object is masked from 'package:MASS':
```

```
##
```

```
##      select
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
library(ggplot2)
```

```
data(aids)
```

```
unique_person = unique(aids$person)
```

```
unique_person
```

```
##      [1] 10002 10005 10029 10039 10048 10052 10079 10088 10092 10131 10132 10135
##      [13] 10145 10171 10173 10175 10191 10196 10204 10213 10221 10222 10259 10263
##      [25] 10273 10290 10302 10304 10323 10343 10344 10350 10360 10361 10362 10372
##      [37] 10388 10396 10401 10403 10416 10419 10424 10425 10432 10433 10437 10444
##      [49] 10453 10473 10526 10527 10538 10557 10564 10569 10579 10587 10591 10642
##      [61] 10662 10669 10675 10678 10700 10770 10773 10806 10865 10878 10915 10916
##      [73] 10949 10954 10956 10968 11005 11048 11062 11076 11088 11100 11106 11118
##      [85] 11131 11142 11143 11165 11172 11175 11199 11200 20003 20013 20014 20032
##      [97] 20042 20057 20066 20072 20082 20086 20089 20111 20143 20147 20158 20175
##     [109] 20199 20205 20232 20240 20284 20305 20323 20324 20332 20344 20348 20362
##     [121] 20363 20374 20393 20395 20397 20404 20417 20421 20439 20476 20477 20492
##     [133] 20498 20523 20537 20567 20568 20571 20583 20584 20591 20595 20604 20605
##     [145] 20616 20664 20678 20713 20723 20736 20748 20749 20768 20776 20777 20779
##     [157] 20837 20839 20850 20851 20852 20891 20906 21029 21058 21083 21087 21090
##     [169] 21093 21136 21194 30007 30010 30018 30024 30038 30046 30048 30049 30050
##     [181] 30051 30069 30075 30083 30101 30119 30122 30132 30133 30135 30148 30173
##     [193] 30177 30179 30183 30193 30216 30225 30239 30262 30281 30301 30306 30310
##     [205] 30324 30372 30376 30382 30388 30392 30405 30412 30420 30428 30454 30485
##     [217] 30489 30490 30498 30503 30504 30508 30515 30531 30536 30548 30562 30599
```

```
## [229] 30654 30663 30673 30677 30692 30693 30698 30699 30702 30713 30735 30777
## [241] 30798 30820 30827 30835 30840 30864 30868 30871 30881 30882 30913 30931
## [253] 30933 30953 30960 30995 30999 31036 31054 31062 40012 40014 40043 40121
## [265] 40132 40175 40217 40224 40249 40286 40327 40337 40340 40362 40363 40372
## [277] 40374 40375 40378 40390 40399 40401 40402 40419 40438 40445 40464 40499
## [289] 40508 40520 40534 40553 40555 40562 40571 40624 40661 40672 40681 40693
## [301] 40702 40738 40759 40774 40791 40795 40807 40867 40873 40901 40904 40942
## [313] 40959 40967 40970 40973 41032 41045 41061 41062 41082 41108 41142 41157
## [325] 41158 41163 41165 41185 41194 41221 41243 41253 41261 41265 41289 41305
## [337] 41314 41325 41328 41395 41402 41406 41407 41411 41414 41416 41452 41474
## [349] 41475 41521 41549 41566 41620 41621 41628 41646 41656 41658 41659 41687
## [361] 41691 41692 41717 41725 41728 41741 41820 41829 41844
```

```
set.seed(666)
randint = sample(unique_person,50)
training_data = aids %>%
  filter(person %in% randint)
training_data
```

##	cd4	time	drugs	partners	packs	cesd	age	person
## 1	561	-1.253936	1	5	0	0	4.40	10145
## 2	1102	-0.755647	1	5	0	0	4.40	10145
## 3	1620	-0.240931	1	-1	0	-7	4.40	10145
## 4	697	0.238193	1	-1	0	-4	4.40	10145
## 5	538	0.766598	1	-4	0	-4	4.40	10145
## 6	811	1.284052	1	-4	0	-5	4.40	10145
## 7	592	1.779603	1	-4	0	-2	4.40	10145
## 8	568	2.277892	1	-3	0	-6	4.40	10145
## 9	384	2.795346	1	-4	0	-3	4.40	10145
## 10	431	3.293634	1	1	0	-7	4.40	10145
## 11	544	3.772758	1	-3	0	-5	4.40	10145
## 12	677	4.290212	1	-3	0	-7	4.40	10145
## 13	587	-1.667351	0	5	0	1	0.63	10304
## 14	638	-1.193703	1	5	0	2	0.63	10304
## 15	583	-0.673511	1	5	0	10	0.63	10304
## 16	479	-0.249144	1	5	0	17	0.63	10304
## 17	588	1.152635	0	5	0	2	0.63	10304
## 18	826	1.401780	1	5	0	7	0.63	10304
## 19	273	1.911020	0	5	0	4	0.63	10304
## 20	324	2.737851	0	-2	0	1	0.63	10304
## 21	515	3.236140	0	0	0	2	0.63	10304
## 22	1472	-1.604381	1	5	4	26	8.78	10343
## 23	1589	-0.254620	1	-1	3	24	8.78	10343
## 24	739	1.519507	1	-3	3	-1	8.78	10343
## 25	1697	-2.726899	1	5	3	14	6.50	10350
## 26	1416	-2.228611	1	5	3	4	6.50	10350
## 27	803	-1.749487	1	5	3	-1	6.50	10350
## 28	713	0.766598	1	1	2	-5	6.50	10350
## 29	599	2.031485	1	-4	2	4	6.50	10350
## 30	556	2.502396	1	-3	2	0	6.50	10350
## 31	668	0.246407	1	5	0	10	3.56	10361
## 32	405	0.744695	1	3	0	9	3.56	10361
## 33	400	1.262149	1	-4	0	5	3.56	10361
## 34	259	1.741273	1	-4	0	7	3.56	10361
## 35	558	2.321697	1	2	0	4	3.56	10361

## 36	488	2.789870	1	5	0	2	3.56	10361
## 37	307	3.288159	1	5	0	0	3.56	10361
## 38	307	3.786448	1	5	0	7	3.56	10361
## 39	326	4.284737	1	5	0	11	3.56	10361
## 40	293	4.802190	1	2	0	24	3.56	10361
## 41	343	5.289528	1	5	0	36	3.56	10361
## 42	631	-1.292266	0	-3	0	-1	-1.12	10473
## 43	558	-0.747433	0	-2	0	6	-1.12	10473
## 44	934	-0.229979	0	-2	0	2	-1.12	10473
## 45	200	0.229979	1	-3	0	8	-1.12	10473
## 46	314	2.338125	0	-4	0	2	-1.12	10473
## 47	149	2.839151	0	-4	0	7	-1.12	10473
## 48	855	-2.699521	1	-3	2	-4	6.20	10587
## 49	1264	-2.193018	1	0	2	-4	6.20	10587
## 50	1259	-1.675565	1	5	1	-5	6.20	10587
## 51	519	-1.256673	1	-3	1	-6	6.20	10587
## 52	1103	0.711841	1	-4	1	-5	6.20	10587
## 53	1082	1.708419	1	-4	1	-4	6.20	10587
## 54	779	2.225873	1	-4	1	-5	6.20	10587
## 55	1245	-1.697467	1	1	0	16	0.36	10669
## 56	1103	-1.199179	1	3	0	11	0.36	10669
## 57	1260	-0.260096	1	3	0	2	0.36	10669
## 58	478	1.316906	1	-5	0	15	0.36	10669
## 59	385	1.891855	1	-4	0	15	0.36	10669
## 60	1236	-2.466804	1	0	0	0	0.36	10773
## 61	942	2.464066	0	0	0	-3	0.36	10773
## 62	950	2.981519	0	-3	0	0	0.36	10773
## 63	1077	-1.248460	1	5	0	11	0.08	10806
## 64	1220	-0.747433	1	3	0	8	0.08	10806
## 65	756	-0.249144	1	5	0	19	0.08	10806
## 66	551	0.249144	0	2	0	28	0.08	10806
## 67	676	0.747433	0	1	0	18	0.08	10806
## 68	1206	-0.240931	1	5	0	1	-0.18	10878
## 69	691	0.240931	0	-2	0	-1	-0.18	10878
## 70	596	-0.758385	1	1	0	8	-3.17	10916
## 71	616	-0.240931	1	-3	0	-3	-3.17	10916
## 72	496	0.240931	1	-1	0	-2	-3.17	10916
## 73	672	0.700890	1	-2	0	-3	-3.17	10916
## 74	359	1.215606	1	-4	0	-2	-3.17	10916
## 75	660	1.771389	1	-2	0	-5	-3.17	10916
## 76	257	2.308008	1	-2	0	-1	-3.17	10916
## 77	383	2.825462	1	-4	0	20	-3.17	10916
## 78	940	-1.097878	0	-2	0	0	23.06	11076
## 79	570	-0.240931	1	0	0	3	23.06	11076
## 80	799	-1.341547	1	-2	1	11	1.79	20014
## 81	913	-0.347707	1	-2	0	3	1.79	20014
## 82	269	0.826831	1	-3	0	11	1.79	20014
## 83	276	1.382615	1	-3	0	1	1.79	20014
## 84	224	1.872690	1	-4	0	6	1.79	20014
## 85	1075	-1.563313	1	-2	0	10	4.51	20032
## 86	865	-0.867899	1	-3	0	4	4.51	20032
## 87	988	-0.720055	1	-3	0	10	4.51	20032
## 88	940	-0.336756	1	0	0	11	4.51	20032
## 89	900	0.336756	1	-1	0	-5	4.51	20032

## 90	556	0.878850	1	-4	0	5	4.51	20032
## 91	754	-2.882957	0	-2	0	-4	13.71	20082
## 92	814	-2.349076	0	3	0	0	13.71	20082
## 93	936	-0.758385	0	5	0	-5	13.71	20082
## 94	796	-0.224504	0	5	0	30	13.71	20082
## 95	770	0.224504	0	4	0	-2	13.71	20082
## 96	676	0.511978	0	1	0	-2	13.71	20082
## 97	811	1.029432	0	5	0	-7	13.71	20082
## 98	662	1.492129	0	0	0	-4	13.71	20082
## 99	780	2.023272	0	5	0	-4	13.71	20082
## 100	857	-2.787132	1	5	0	-4	4.42	20240
## 101	375	-2.422998	1	5	0	3	4.42	20240
## 102	543	-2.078029	1	5	0	-3	4.42	20240
## 103	530	-1.382615	1	5	0	10	4.42	20240
## 104	525	-0.900753	1	5	0	5	4.42	20240
## 105	633	-0.251882	1	5	0	17	4.42	20240
## 106	908	-0.262834	0	5	0	-6	-2.37	20393
## 107	250	0.262834	1	3	0	-7	-2.37	20393
## 108	307	0.818617	1	3	0	-6	-2.37	20393
## 109	605	1.330595	1	-1	0	-3	-2.37	20393
## 110	546	1.796030	1	0	0	-7	-2.37	20393
## 111	372	2.255989	1	-3	0	-4	-2.37	20393
## 112	339	2.852840	1	1	0	-7	-2.37	20393
## 113	311	3.175907	1	1	0	-2	-2.37	20393
## 114	319	3.466119	1	1	0	-7	-2.37	20393
## 115	492	4.000000	1	0	0	-4	-2.37	20393
## 116	350	4.536619	1	-1	0	-2	-2.37	20393
## 117	442	5.103354	1	5	0	2	-2.37	20393
## 118	838	-2.603696	1	-1	0	-2	-1.35	20404
## 119	412	-1.555099	1	-1	0	0	-1.35	20404
## 120	888	-0.678987	1	0	1	-6	-1.35	20404
## 121	724	-0.268309	1	1	1	-5	-1.35	20404
## 122	721	0.268309	1	-2	1	-7	-1.35	20404
## 123	690	-0.506502	1	5	0	-5	-7.76	20417
## 124	1025	0.511978	1	-3	0	-5	-7.76	20417
## 125	483	0.966461	1	-4	0	1	-7.76	20417
## 126	935	-0.763860	1	0	0	3	3.35	20421
## 127	816	-0.260096	1	-2	0	3	3.35	20421
## 128	1119	0.251882	1	-1	0	0	3.35	20421
## 129	902	0.799452	1	-4	0	-3	3.35	20421
## 130	809	1.286790	1	-4	0	3	3.35	20421
## 131	374	1.785079	1	-4	0	3	3.35	20421
## 132	684	2.321697	1	-4	0	3	3.35	20421
## 133	725	3.687885	0	-4	0	24	3.35	20421
## 134	892	-2.776181	1	0	0	-3	-9.52	20477
## 135	1110	-2.280630	1	0	0	4	-9.52	20477
## 136	740	-0.117728	1	-2	0	-4	-9.52	20477
## 137	472	0.117728	1	-3	0	2	-9.52	20477
## 138	755	0.481862	1	1	0	-1	-9.52	20477
## 139	701	1.018481	1	2	0	1	-9.52	20477
## 140	37	1.555099	1	5	0	-1	-9.52	20477
## 141	470	2.072553	1	1	0	16	-9.52	20477
## 142	957	-1.839836	1	-2	1	-7	3.36	20567
## 143	1006	-1.234771	1	-4	0	-3	3.36	20567

## 144	936	-0.802190	1	-4	0	-7	3.36	20567
## 145	492	0.717317	1	-4	0	-7	3.36	20567
## 146	1629	1.314168	1	-4	0	-7	3.36	20567
## 147	1203	1.615332	1	-4	0	3	3.36	20567
## 148	1397	2.439425	1	-4	0	-1	3.36	20567
## 149	424	-1.535934	1	-1	0	11	20.66	20571
## 150	553	-1.152635	1	-1	0	8	20.66	20571
## 151	734	-0.769336	1	-1	0	11	20.66	20571
## 152	405	-0.328542	1	0	0	9	20.66	20571
## 153	596	0.328542	1	1	0	8	20.66	20571
## 154	396	0.884326	1	-5	0	11	20.66	20571
## 155	831	-1.259411	1	0	2	4	-1.57	20584
## 156	753	-0.761123	1	3	2	-6	-1.57	20584
## 157	508	-0.260096	1	-3	2	3	-1.57	20584
## 158	406	1.366188	1	-4	1	-7	-1.57	20584
## 159	175	1.809719	1	-4	2	-6	-1.57	20584
## 160	108	2.598220	1	-4	2	-7	-1.57	20584
## 161	1292	-1.459274	1	5	3	-7	0.46	20595
## 162	884	-1.040383	1	5	3	-7	0.46	20595
## 163	785	-0.747433	1	5	3	-7	0.46	20595
## 164	1243	-0.268309	1	5	3	-7	0.46	20595
## 165	789	0.268309	1	5	3	-5	0.46	20595
## 166	517	0.769336	1	5	0	0	0.46	20595
## 167	1022	-0.281999	1	-4	2	2	-6.90	20678
## 168	800	-1.270363	1	5	0	-2	4.47	20713
## 169	743	-0.763860	1	5	0	14	4.47	20713
## 170	719	-0.251882	1	-2	0	11	4.47	20713
## 171	481	1.223819	0	-4	0	10	4.47	20713
## 172	348	1.971253	0	-4	0	-7	4.47	20713
## 173	392	2.362765	0	-4	0	-6	4.47	20713
## 174	1451	-0.254620	1	5	0	-3	1.93	20749
## 175	1684	0.249144	1	5	0	0	1.93	20749
## 176	797	0.744695	1	1	0	-6	1.93	20749
## 177	916	1.289528	1	3	0	-2	1.93	20749
## 178	918	1.897331	1	4	0	-2	1.93	20749
## 179	1345	2.247776	1	2	0	-4	1.93	20749
## 180	821	3.110198	1	5	0	-5	1.93	20749
## 181	759	-0.210815	1	5	1	12	4.37	20852
## 182	969	0.210815	1	5	1	6	4.37	20852
## 183	664	1.171800	1	2	1	14	4.37	20852
## 184	587	1.664613	1	1	1	12	4.37	20852
## 185	499	2.469541	1	5	1	17	4.37	20852
## 186	486	3.011636	1	-4	1	10	4.37	20852
## 187	805	3.493498	1	-4	1	9	4.37	20852
## 188	400	3.972621	1	-1	1	18	4.37	20852
## 189	567	4.583162	1	-1	1	7	4.37	20852
## 190	926	-0.221766	0	5	0	-2	29.08	20891
## 191	354	0.221766	1	-2	0	0	29.08	20891
## 192	899	0.550308	0	-4	0	-3	29.08	20891
## 193	675	1.081451	0	-4	0	-2	29.08	20891
## 194	524	1.585216	0	-4	0	-6	29.08	20891
## 195	1082	2.045175	0	-4	0	-7	29.08	20891
## 196	590	2.521561	0	-4	0	-7	29.08	20891
## 197	861	2.825462	0	-4	0	-7	29.08	20891

## 198	786	3.252567	0	-4	0	-6	29.08	20891
## 199	728	3.822040	0	-4	0	-4	29.08	20891
## 200	764	4.262834	0	-4	0	-6	29.08	20891
## 201	768	-0.832307	1	0	3	0	-6.44	30075
## 202	656	-0.238193	1	5	3	4	-6.44	30075
## 203	511	0.238193	1	-4	3	12	-6.44	30075
## 204	710	0.698152	1	-4	4	21	-6.44	30075
## 205	641	1.070500	1	-4	3	3	-6.44	30075
## 206	916	2.425736	1	-3	4	0	-6.44	30075
## 207	875	-1.549623	1	0	0	2	4.17	30083
## 208	972	-0.621492	1	5	0	-1	4.17	30083
## 209	609	0.621492	1	-3	0	10	4.17	30083
## 210	504	-2.532512	1	5	0	-2	-0.62	30132
## 211	695	-2.110883	1	-1	0	8	-0.62	30132
## 212	900	-1.708419	1	4	0	0	-0.62	30132
## 213	665	-1.248460	1	0	0	0	-0.62	30132
## 214	1166	-0.249144	1	5	0	10	-0.62	30132
## 215	274	0.249144	1	0	0	-1	-0.62	30132
## 216	995	-0.249144	1	-3	3	-7	-2.53	30376
## 217	975	0.249144	1	-4	3	0	-2.53	30376
## 218	782	0.824093	1	-2	3	0	-2.53	30376
## 219	2534	-2.433949	1	5	4	25	13.25	30503
## 220	394	2.433949	1	5	3	8	13.25	30503
## 221	748	-2.466804	1	5	0	25	2.55	30735
## 222	685	-1.968515	1	5	0	21	2.55	30735
## 223	684	-1.546886	1	5	0	19	2.55	30735
## 224	711	-1.144422	1	5	0	25	2.55	30735
## 225	837	-0.703628	1	5	0	22	2.55	30735
## 226	897	-0.224504	1	5	0	17	2.55	30735
## 227	606	0.435318	1	0	0	22	2.55	30735
## 228	431	0.952772	1	1	0	22	2.55	30735
## 229	473	1.467488	1	5	0	18	2.55	30735
## 230	370	1.965777	1	5	0	16	2.55	30735
## 231	303	-0.249144	0	5	0	17	11.53	30827
## 232	451	0.249144	1	0	0	-5	11.53	30827
## 233	297	0.673511	1	-2	0	17	11.53	30827
## 234	233	1.040383	0	-3	0	2	11.53	30827
## 235	290	1.620808	0	-2	0	-1	11.53	30827
## 236	184	1.908282	0	-4	0	-4	11.53	30827
## 237	505	2.417522	0	-2	0	-6	11.53	30827
## 238	301	2.915811	0	-3	0	-4	11.53	30827
## 239	101	3.436003	0	-3	0	-5	11.53	30827
## 240	1585	-0.763860	0	0	3	15	-6.17	31054
## 241	1731	-0.227242	1	1	3	16	-6.17	31054
## 242	1466	0.227242	1	5	3	9	-6.17	31054
## 243	1224	0.610541	1	0	3	1	-6.17	31054
## 244	1482	1.108830	1	-4	3	0	-6.17	31054
## 245	1376	1.694730	1	-4	0	-3	-6.17	31054
## 246	1245	2.088980	0	-4	0	-2	-6.17	31054
## 247	1159	2.748802	1	-4	3	-6	-6.17	31054
## 248	916	3.400411	1	-4	3	-5	-6.17	31054
## 249	803	-0.325804	1	5	4	1	11.22	40224
## 250	664	0.325804	1	0	3	1	11.22	40224
## 251	539	0.804928	0	-2	3	-4	11.22	40224

##	252	311	1.648186	0	-4	3	1	11.22	40224
##	253	735	-1.694730	1	5	0	9	7.55	40372
##	254	966	-1.067762	1	5	0	-7	7.55	40372
##	255	772	-0.665298	1	1	0	-3	7.55	40372
##	256	749	-0.260096	1	-1	0	-6	7.55	40372
##	257	423	1.979466	1	-3	0	8	7.55	40372
##	258	446	2.554415	1	-4	0	0	7.55	40372
##	259	605	3.088296	1	0	0	-5	7.55	40372
##	260	534	3.493498	1	0	0	-2	7.55	40372
##	261	387	4.062971	1	-2	0	5	7.55	40372
##	262	903	-1.700205	1	5	3	-5	5.74	40402
##	263	426	2.198494	1	-4	2	-3	5.74	40402
##	264	942	-2.910335	1	5	3	-7	11.59	40499
##	265	699	-2.422998	1	5	3	-5	11.59	40499
##	266	907	-1.957563	1	4	3	-7	11.59	40499
##	267	734	-1.552361	1	-2	3	-7	11.59	40499
##	268	1000	-0.295688	0	0	3	13	0.71	40738
##	269	480	0.295688	1	2	3	-1	0.71	40738
##	270	449	0.835045	1	-2	3	8	0.71	40738
##	271	407	1.448323	1	-2	3	22	0.71	40738
##	272	360	1.946612	1	-3	2	5	0.71	40738
##	273	325	3.457906	1	-5	2	1	0.71	40738
##	274	382	4.881588	1	-4	2	-3	0.71	40738
##	275	882	-0.334018	1	5	3	3	18.14	40774
##	276	713	0.334018	1	0	0	3	18.14	40774
##	277	512	0.835045	1	-2	3	3	18.14	40774
##	278	389	1.349760	1	-4	3	4	18.14	40774
##	279	1157	-2.521561	1	2	4	33	-4.55	40795
##	280	972	-2.099931	1	5	4	33	-4.55	40795
##	281	1350	-1.661875	1	5	4	39	-4.55	40795
##	282	1290	-1.182752	1	5	4	39	-4.55	40795
##	283	1246	-0.240931	1	5	4	48	-4.55	40795
##	284	530	0.240931	1	5	4	48	-4.55	40795
##	285	546	1.259411	1	5	4	40	-4.55	40795
##	286	394	2.291581	1	5	4	39	-4.55	40795
##	287	906	-1.853525	1	5	0	-1	15.23	41062
##	288	1212	-1.125257	1	5	0	-6	15.23	41062
##	289	1459	-0.703628	1	5	0	-1	15.23	41062
##	290	1538	-0.235455	1	0	0	-7	15.23	41062
##	291	1362	0.235455	1	0	0	-5	15.23	41062
##	292	624	1.396304	1	-2	0	-5	15.23	41062
##	293	170	2.354552	1	-4	0	-5	15.23	41062
##	294	277	2.819986	1	-5	0	-2	15.23	41062
##	295	980	-0.821355	1	-3	0	0	7.67	41158
##	296	1174	-0.251882	1	1	0	-7	7.67	41158
##	297	1008	0.251882	1	-1	0	-4	7.67	41158
##	298	931	0.750171	1	-2	0	-6	7.67	41158
##	299	910	1.763176	1	-2	0	-3	7.67	41158
##	300	327	2.702259	1	-1	0	-4	7.67	41158
##	301	295	3.200547	1	-4	0	-1	7.67	41158
##	302	83	3.698836	1	-3	0	-2	7.67	41158
##	303	819	-2.647502	0	5	0	1	0.67	41474
##	304	970	-2.242300	0	5	0	0	0.67	41474
##	305	647	-1.760438	0	5	0	2	0.67	41474

```
## 306 828 -1.264887 0 -2 0 4 0.67 41474
## 307 647 0.268309 0 -3 0 6 0.67 41474
## 308 624 0.783025 0 -4 0 8 0.67 41474
## 309 547 1.347023 0 5 0 8 0.67 41474
## 310 628 1.768652 0 1 0 2 0.67 41474
## 311 532 -1.779603 1 5 0 21 2.49 41691
## 312 464 -1.284052 1 5 0 16 2.49 41691
## 313 780 -0.766598 1 3 0 33 2.49 41691
## 314 633 -0.369610 1 5 0 31 2.49 41691
## 315 540 1.377139 1 5 0 38 2.49 41691
```

```
bandwidths <- seq(15, 40, length.out = 100)
```

```
x = training_data$time
y = training_data$cd4
smoother = ksmooth(x,y, kernel = "normal", bandwidth = 2, x.points = unique(x))
smoothed_values = smoother$y
  smoothed_full = smoothed_values[match(x, unique(x))]
  residuals = y - smoothed_full
noise_variance = var(residuals)
noise_variance
```

```
## [1] 142819.3
```

```
#b
```

```
cp_values <- numeric(length(bandwidths))
for (i in 1:length(bandwidths)) {
  bandwidth <- bandwidths[i]
  smoother = ksmooth(x,y, kernel = "normal", bandwidth = bandwidth, x.points = unique(x))
  smoothed_values = smoother$y
  smoothed_full = smoothed_values[match(x, unique(x))]
  residuals = y - smoothed_full
  rss = sum(residuals^2)
  noise_variance = var(residuals)
  n = length(x)
  kernel_weights <- exp(-0.5 * (outer(x, x, "-") / bandwidth)^2)
  smoother_matrix_diag <- diag(kernel_weights)
  trace_S_lambda = sum(smoother_matrix_diag)
  cp = 1/n * (rss + 2 * noise_variance * trace_S_lambda)
  cp_values[i] <- cp
}
```

```
cp_values
```

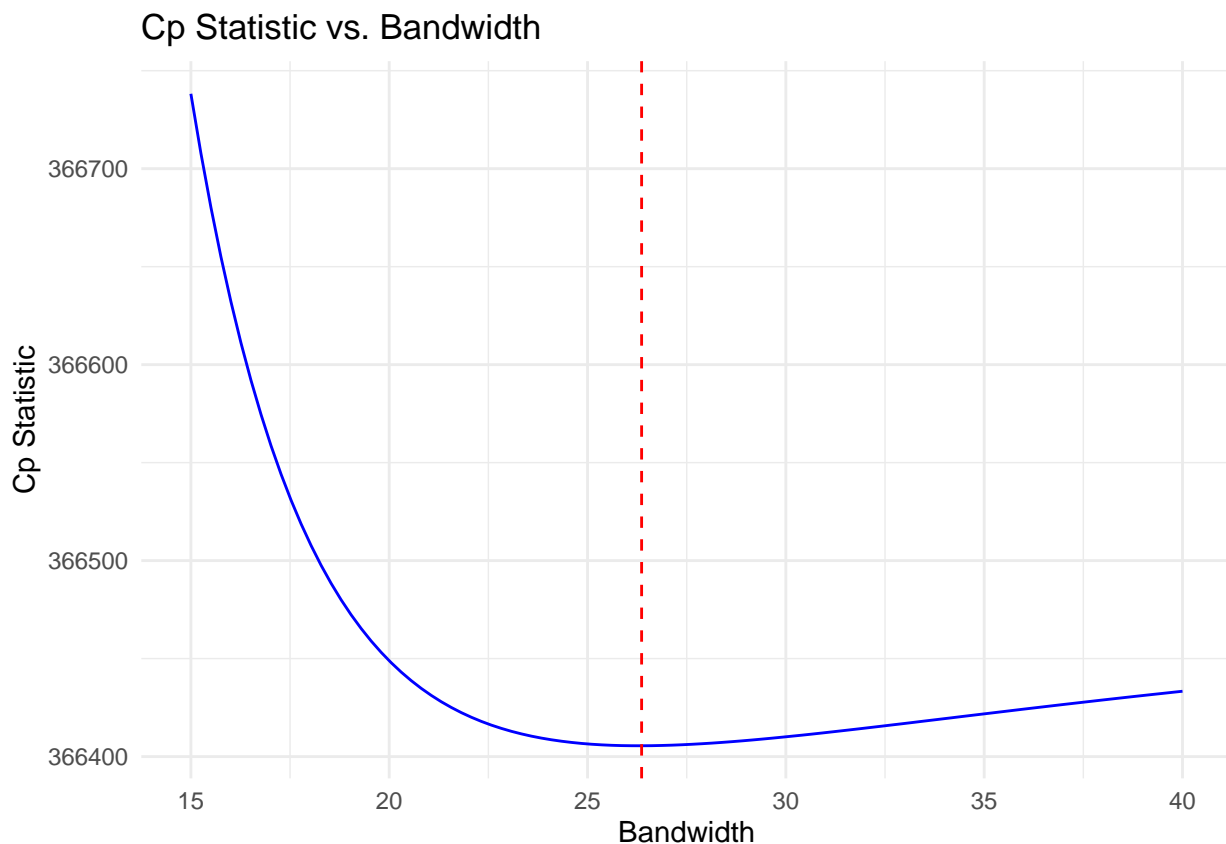
```
## [1] 366738.2 366707.9 366680.2 366655.0 366632.1 366611.1 366592.0 366574.6
## [9] 366558.6 366544.1 366530.8 366518.7 366507.7 366497.6 366488.4 366480.0
## [17] 366472.3 366465.3 366459.0 366453.2 366447.9 366443.1 366438.7 366434.8
## [25] 366431.2 366428.0 366425.1 366422.4 366420.1 366417.9 366416.1 366414.4
## [33] 366412.9 366411.6 366410.4 366409.4 366408.6 366407.8 366407.2 366406.7
## [41] 366406.3 366406.0 366405.8 366405.6 366405.5 366405.5 366405.6 366405.7
## [49] 366405.8 366406.0 366406.3 366406.5 366406.9 366407.2 366407.6 366408.0
## [57] 366408.4 366408.9 366409.4 366409.9 366410.4 366410.9 366411.5 366412.0
## [65] 366412.6 366413.2 366413.7 366414.3 366414.9 366415.5 366416.1 366416.8
## [73] 366417.4 366418.0 366418.6 366419.2 366419.8 366420.4 366421.1 366421.7
## [81] 366422.3 366422.9 366423.5 366424.1 366424.7 366425.3 366425.9 366426.5
## [89] 366427.1 366427.7 366428.3 366428.9 366429.4 366430.0 366430.6 366431.1
```



```
## [97] 366431.7 366432.2 366432.8 366433.3
optimal_bandwidth_idx <- which.min(cp_values)
optimal_bandwidth <- bandwidths[optimal_bandwidth_idx]
smoother = ksmooth(x,y, kernel = "normal", bandwidth = optimal_bandwidth, x.points = unique(x))
smoothed_values = smoother$y
smoothed_full = smoothed_values[match(x, unique(x))]
residuals = y - smoothed_full
rss = sum(residuals^2)
training_error = rss/length(x)
training_error
```

```
## [1] 121876.9
cp_plot <- ggplot(data.frame(bandwidths, cp_values), aes(x = bandwidths, y = cp_values)) +
  geom_line(color = "blue") +
  geom_vline(xintercept = optimal_bandwidth, color = "red", linetype = "dashed") +
  labs(title = "Cp Statistic vs. Bandwidth", x = "Bandwidth", y = "Cp Statistic") +
  theme_minimal()

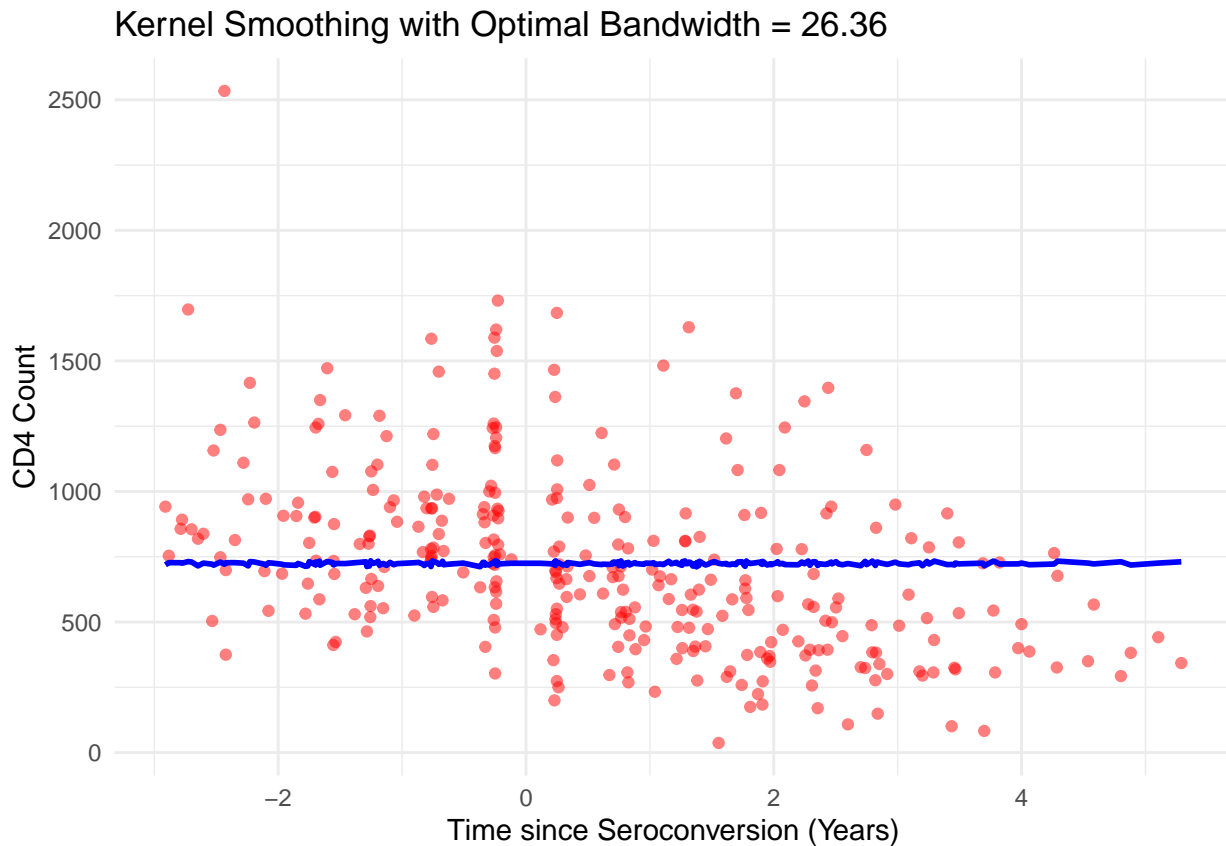
print(cp_plot)
```



```
fitted_plot <- ggplot(data.frame(x, y), aes(x = x, y = y)) +
  geom_point(color = "red", alpha = 0.5) +
  geom_line(aes(x = x, y = smoothed_full), color = "blue", size = 1) +
  labs(title = paste("Kernel Smoothing with Optimal Bandwidth =", round(optimal_bandwidth, 2)),
       x = "Time since Seroconversion (Years)", y = "CD4 Count") +
  theme_minimal()
```

```
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

```
print(fitted_plot)
```



```
#c
```

```
bootstrapped_loo <- function(x, y, bandwidth, n_bootstrap = 100) {
```

```
  n <- length(x)
  bootstrap_errors <- numeric(n_bootstrap)
  for (i in 1:n_bootstrap) {
    bootstrap_indices <- sample(1:n, size = n, replace = TRUE)
    x_bootstrap <- x[bootstrap_indices]
    y_bootstrap <- y[bootstrap_indices]
    loo_errors <- numeric(n)
    for (j in 1:n) {
      x_train <- x_bootstrap[-j]
      y_train <- y_bootstrap[-j]
      smoother_train <- ksmooth(x_train, y_train, kernel = "normal", bandwidth = bandwidth, x.points = 1)
      smoothed_train_values <- smoother_train$y
      prediction <- smoothed_train_values[match(x[j], unique(x))]
      loo_errors[j] <- (y[j] - prediction)^2
    }
    bootstrap_errors[i] <- mean(loo_errors)
  }
}
```

```

    mean_bootstrap_error <- mean(bootstrap_errors)
    return(mean_bootstrap_error)
}

n_bootstrap <- 100

bootstrap_error <- bootstrapped_loo(x, y, optimal_bandwidth, n_bootstrap)

print(paste("Bootstrap Estimated Prediction Error:", bootstrap_error))

## [1] "Bootstrap Estimated Prediction Error: 122290.173021034"

#d
estimator_632 = 0.368 * training_error + 0.632 * bootstrap_error
estimator_632

## [1] 122138.1
training_error

## [1] 121876.9
bootstrap_error

## [1] 122290.2

We can tell that the .632 estimator is living between training_error and bootstrap_error

#e
testing_data = aids %>%
  filter(!(person %in% randint))
unique_person = unique(testing_data$person)

mean_errors = numeric(1000)
compute_mean_error <- function(x_test, y_test, bandwidth) {
  smoothed_values <- smoother$y[match(x_test, unique(x_test))]
  mean_squared_error <- mean((y_test - smoothed_values)^2)
  return(mean_squared_error)
}

smoother = ksmooth(x,y, kernel = "normal", bandwidth = optimal_bandwidth, x.points = unique(x))
for (i in 1:1000) {
  randint = sample(unique_person,5)
  testing = testing_data %>%
    filter(person %in% randint)
  x_test_sample <- testing$time
  y_test_sample <- testing$cd4
  mean_errors[i] <- compute_mean_error(x_test_sample, y_test_sample, optimal_bandwidth)
}

average_mean_squared_error = mean(mean_errors)
average_mean_squared_error

## [1] 164957.4

```

The prediction error is a bit higher than the training error but acceptable.