

# Análisis y Tratamiento de Datos con R: Departamento de Matemática

\*

6 de noviembre de 2017

## Imputación de datos

Basados en: <https://www.r-bloggers.com/missing-value-treatment/>

```
# initialize the data
data ("BostonHousing", package="mlbench")
original <- BostonHousing # backup original data

# Introduce missing values
set.seed(100)
BostonHousing[sample(1:nrow(BostonHousing), 40), "rad"] <- NA
BostonHousing[sample(1:nrow(BostonHousing), 40), "ptratio"] <- NA
```

¿Que datos están perdidos?

```
# Pattern of missing values
library(mice)
```

```
## Loading required package: lattice
```

```
md.pattern(BostonHousing)
```

```
##      crim zn  indus chas nox  rm  age  dis  tax  b  lstat medv rad ptratio
## 431    1  1    1    1  1  1  1  1  1  1  1  1  1  1  0
## 35    1  1    1    1  1  1  1  1  1  1  1  1  0  1  1
## 35    1  1    1    1  1  1  1  1  1  1  1  1  1  0  1
## 5     1  1    1    1  1  1  1  1  1  1  1  1  0  0  2
##      0  0    0    0  0  0  0  0  0  0  0  0  0  40  80
```

## Borrando observaciones

- Si se tiene suficientes observaciones, no se debería perder mucho
- No se introduce sesgos debido a alguna pérdida sistemática de datos (representatividad)

```
# Example
lm(medv ~ ptratio + rad, data=BostonHousing, na.action=na.omit)
```

---

\*,

```
##
## Call:
## lm(formula = medv ~ ptratio + rad, data = BostonHousing, na.action = na.omit)
##
## Coefficients:
## (Intercept)      ptratio          rad
##      57.2724      -1.7836      -0.2035
```

## Borrando variables

Depende de la importancia del poder predictor que pueda tener esa variable.

## Imputación usando la media, mediana o moda

```
library(Hmisc)
```

```
## Loading required package: survival
```

```
## Loading required package: Formula
```

```
## Loading required package: ggplot2
```

```
##
```

```
## Attaching package: 'Hmisc'
```

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

```
##
```

```
##      format.pval, round.POSIXt, trunc.POSIXt, units
```

```
impute(BostonHousing$ptratio, mean) # replace with mean
```

```
##      1      2      3      4      5      6      7      8
## 15.3000 17.8000 17.8000 18.7000 18.7000 18.7000 15.2000 15.2000
##      9     10     11     12     13     14     15     16
## 15.2000 15.2000 15.2000 15.2000 15.2000 21.0000 21.0000 21.0000
##     17     18     19     20     21     22     23     24
## 21.0000 21.0000 21.0000 21.0000 21.0000 21.0000 21.0000 21.0000
##     25     26     27     28     29     30     31     32
## 21.0000 21.0000 21.0000 21.0000 21.0000 21.0000 21.0000 21.0000
##     33     34     35     36     37     38     39     40
## 21.0000 21.0000 21.0000 19.2000 19.2000 19.2000 19.2000 18.3000
##     41     42     43     44     45     46     47     48
## 18.3000 17.9000 18.4676* 17.9000 17.9000 17.9000 17.9000 17.9000
##     49     50     51     52     53     54     55     56
## 17.9000 17.9000 16.8000 16.8000 16.8000 16.8000 21.1000 17.9000
##     57     58     59     60     61     62     63     64
## 17.3000 15.1000 19.7000 19.7000 18.4676* 19.7000 19.7000 19.7000
##     65     66     67     68     69     70     71     72
## 18.6000 16.1000 16.1000 18.9000 18.9000 18.9000 19.2000 19.2000
##     73     74     75     76     77     78     79     80
## 19.2000 19.2000 18.7000 18.7000 18.7000 18.7000 18.7000 18.7000
##     81     82     83     84     85     86     87     88
## 19.0000 19.0000 19.0000 19.0000 18.5000 18.5000 18.5000 18.5000
##     89     90     91     92     93     94     95     96
```

##	17.8000	17.8000	17.8000	17.8000	18.2000	18.2000	18.2000	18.0000
##	97	98	99	100	101	102	103	104
##	18.0000	18.0000	18.4676*	18.0000	20.9000	20.9000	18.4676*	18.4676*
##	105	106	107	108	109	110	111	112
##	20.9000	20.9000	20.9000	20.9000	20.9000	20.9000	20.9000	17.8000
##	113	114	115	116	117	118	119	120
##	18.4676*	17.8000	17.8000	17.8000	18.4676*	18.4676*	17.8000	17.8000
##	121	122	123	124	125	126	127	128
##	19.1000	19.1000	19.1000	19.1000	18.4676*	19.1000	19.1000	21.2000
##	129	130	131	132	133	134	135	136
##	21.2000	21.2000	21.2000	21.2000	21.2000	21.2000	21.2000	18.4676*
##	137	138	139	140	141	142	143	144
##	21.2000	21.2000	21.2000	21.2000	21.2000	21.2000	14.7000	14.7000
##	145	146	147	148	149	150	151	152
##	14.7000	14.7000	14.7000	14.7000	14.7000	14.7000	14.7000	14.7000
##	153	154	155	156	157	158	159	160
##	18.4676*	14.7000	14.7000	18.4676*	14.7000	14.7000	14.7000	14.7000
##	161	162	163	164	165	166	167	168
##	14.7000	14.7000	14.7000	18.4676*	14.7000	14.7000	14.7000	18.4676*
##	169	170	171	172	173	174	175	176
##	14.7000	14.7000	14.7000	14.7000	18.4676*	16.6000	16.6000	16.6000
##	177	178	179	180	181	182	183	184
##	16.6000	16.6000	16.6000	17.8000	17.8000	17.8000	17.8000	17.8000
##	185	186	187	188	189	190	191	192
##	17.8000	17.8000	17.8000	15.2000	15.2000	15.2000	15.2000	15.2000
##	193	194	195	196	197	198	199	200
##	15.2000	15.6000	15.6000	14.4000	18.4676*	12.6000	12.6000	17.0000
##	201	202	203	204	205	206	207	208
##	17.0000	14.7000	14.7000	14.7000	14.7000	18.6000	18.6000	18.6000
##	209	210	211	212	213	214	215	216
##	18.6000	18.6000	18.6000	18.6000	18.6000	18.4676*	18.4676*	18.6000
##	217	218	219	220	221	222	223	224
##	16.4000	16.4000	18.4676*	16.4000	17.4000	17.4000	17.4000	17.4000
##	225	226	227	228	229	230	231	232
##	17.4000	18.4676*	17.4000	17.4000	17.4000	17.4000	17.4000	17.4000
##	233	234	235	236	237	238	239	240
##	17.4000	17.4000	17.4000	17.4000	17.4000	17.4000	16.6000	16.6000
##	241	242	243	244	245	246	247	248
##	16.6000	16.6000	16.6000	16.6000	19.1000	19.1000	18.4676*	19.1000
##	249	250	251	252	253	254	255	256
##	19.1000	19.1000	19.1000	19.1000	19.1000	19.1000	16.4000	16.4000
##	257	258	259	260	261	262	263	264
##	15.9000	13.0000	13.0000	13.0000	13.0000	13.0000	13.0000	13.0000
##	265	266	267	268	269	270	271	272
##	13.0000	13.0000	13.0000	13.0000	13.0000	18.6000	18.6000	18.4676*
##	273	274	275	276	277	278	279	280
##	18.6000	18.6000	17.6000	17.6000	17.6000	17.6000	17.6000	14.9000
##	281	282	283	284	285	286	287	288
##	14.9000	14.9000	14.9000	13.6000	15.3000	15.3000	18.2000	16.6000
##	289	290	291	292	293	294	295	296
##	16.6000	16.6000	18.4676*	18.4676*	19.2000	16.0000	18.4676*	16.0000
##	297	298	299	300	301	302	303	304
##	16.0000	16.0000	14.8000	14.8000	14.8000	16.1000	18.4676*	16.1000
##	305	306	307	308	309	310	311	312

##	18.4000	18.4000	18.4000	18.4000	18.4000	18.4000	18.4000	18.4000
##	313	314	315	316	317	318	319	320
##	18.4676*	18.4676*	18.4000	18.4000	18.4000	18.4000	18.4000	18.4000
##	321	322	323	324	325	326	327	328
##	19.6000	19.6000	19.6000	19.6000	19.6000	19.6000	18.4676*	19.6000
##	329	330	331	332	333	334	335	336
##	16.9000	16.9000	16.9000	18.4676*	16.9000	20.2000	20.2000	20.2000
##	337	338	339	340	341	342	343	344
##	20.2000	20.2000	20.2000	20.2000	20.2000	15.5000	15.9000	17.6000
##	345	346	347	348	349	350	351	352
##	17.6000	18.8000	18.8000	17.9000	17.0000	19.7000	19.7000	18.3000
##	353	354	355	356	357	358	359	360
##	18.3000	17.0000	22.0000	22.0000	20.2000	20.2000	20.2000	20.2000
##	361	362	363	364	365	366	367	368
##	20.2000	20.2000	20.2000	18.4676*	20.2000	20.2000	20.2000	20.2000
##	369	370	371	372	373	374	375	376
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	377	378	379	380	381	382	383	384
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	385	386	387	388	389	390	391	392
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	18.4676*	18.4676*
##	393	394	395	396	397	398	399	400
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	401	402	403	404	405	406	407	408
##	20.2000	20.2000	18.4676*	20.2000	20.2000	20.2000	20.2000	20.2000
##	409	410	411	412	413	414	415	416
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	417	418	419	420	421	422	423	424
##	18.4676*	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	425	426	427	428	429	430	431	432
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	433	434	435	436	437	438	439	440
##	20.2000	20.2000	20.2000	20.2000	18.4676*	20.2000	20.2000	20.2000
##	441	442	443	444	445	446	447	448
##	20.2000	18.4676*	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	449	450	451	452	453	454	455	456
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	457	458	459	460	461	462	463	464
##	20.2000	18.4676*	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	465	466	467	468	469	470	471	472
##	18.4676*	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	473	474	475	476	477	478	479	480
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	481	482	483	484	485	486	487	488
##	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000	20.2000
##	489	490	491	492	493	494	495	496
##	20.1000	20.1000	20.1000	20.1000	20.1000	19.2000	19.2000	19.2000
##	497	498	499	500	501	502	503	504
##	19.2000	19.2000	19.2000	18.4676*	19.2000	21.0000	21.0000	21.0000
##	505	506						
##	21.0000	21.0000						

```
impute(BostonHousing$ptratio, median) # median
```

##	1	2	3	4	5	6	7	8	9	10	11	12
##	15.3	17.8	17.8	18.7	18.7	18.7	15.2	15.2	15.2	15.2	15.2	15.2
##	13	14	15	16	17	18	19	20	21	22	23	24
##	15.2	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0
##	25	26	27	28	29	30	31	32	33	34	35	36
##	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	21.0	19.2
##	37	38	39	40	41	42	43	44	45	46	47	48
##	19.2	19.2	19.2	18.3	18.3	17.9	19.1*	17.9	17.9	17.9	17.9	17.9
##	49	50	51	52	53	54	55	56	57	58	59	60
##	17.9	17.9	16.8	16.8	16.8	16.8	21.1	17.9	17.3	15.1	19.7	19.7
##	61	62	63	64	65	66	67	68	69	70	71	72
##	19.1*	19.7	19.7	19.7	18.6	16.1	16.1	18.9	18.9	18.9	19.2	19.2
##	73	74	75	76	77	78	79	80	81	82	83	84
##	19.2	19.2	18.7	18.7	18.7	18.7	18.7	18.7	19.0	19.0	19.0	19.0
##	85	86	87	88	89	90	91	92	93	94	95	96
##	18.5	18.5	18.5	18.5	17.8	17.8	17.8	17.8	18.2	18.2	18.2	18.0
##	97	98	99	100	101	102	103	104	105	106	107	108
##	18.0	18.0	19.1*	18.0	20.9	20.9	19.1*	19.1*	20.9	20.9	20.9	20.9
##	109	110	111	112	113	114	115	116	117	118	119	120
##	20.9	20.9	20.9	17.8	19.1*	17.8	17.8	17.8	19.1*	19.1*	17.8	17.8
##	121	122	123	124	125	126	127	128	129	130	131	132
##	19.1	19.1	19.1	19.1	19.1*	19.1	19.1	21.2	21.2	21.2	21.2	21.2
##	133	134	135	136	137	138	139	140	141	142	143	144
##	21.2	21.2	21.2	19.1*	21.2	21.2	21.2	21.2	21.2	21.2	14.7	14.7
##	145	146	147	148	149	150	151	152	153	154	155	156
##	14.7	14.7	14.7	14.7	14.7	14.7	14.7	14.7	19.1*	14.7	14.7	19.1*
##	157	158	159	160	161	162	163	164	165	166	167	168
##	14.7	14.7	14.7	14.7	14.7	14.7	14.7	19.1*	14.7	14.7	14.7	19.1*
##	169	170	171	172	173	174	175	176	177	178	179	180
##	14.7	14.7	14.7	14.7	19.1*	16.6	16.6	16.6	16.6	16.6	16.6	17.8
##	181	182	183	184	185	186	187	188	189	190	191	192
##	17.8	17.8	17.8	17.8	17.8	17.8	17.8	15.2	15.2	15.2	15.2	15.2
##	193	194	195	196	197	198	199	200	201	202	203	204
##	15.2	15.6	15.6	14.4	19.1*	12.6	12.6	17.0	17.0	14.7	14.7	14.7
##	205	206	207	208	209	210	211	212	213	214	215	216
##	14.7	18.6	18.6	18.6	18.6	18.6	18.6	18.6	18.6	19.1*	19.1*	18.6
##	217	218	219	220	221	222	223	224	225	226	227	228
##	16.4	16.4	19.1*	16.4	17.4	17.4	17.4	17.4	17.4	19.1*	17.4	17.4
##	229	230	231	232	233	234	235	236	237	238	239	240
##	17.4	17.4	17.4	17.4	17.4	17.4	17.4	17.4	17.4	17.4	16.6	16.6
##	241	242	243	244	245	246	247	248	249	250	251	252
##	16.6	16.6	16.6	16.6	19.1	19.1	19.1*	19.1	19.1	19.1	19.1	19.1
##	253	254	255	256	257	258	259	260	261	262	263	264
##	19.1	19.1	16.4	16.4	15.9	13.0	13.0	13.0	13.0	13.0	13.0	13.0
##	265	266	267	268	269	270	271	272	273	274	275	276
##	13.0	13.0	13.0	13.0	13.0	18.6	18.6	19.1*	18.6	18.6	17.6	17.6
##	277	278	279	280	281	282	283	284	285	286	287	288
##	17.6	17.6	17.6	14.9	14.9	14.9	14.9	13.6	15.3	15.3	18.2	16.6
##	289	290	291	292	293	294	295	296	297	298	299	300
##	16.6	16.6	19.1*	19.1*	19.2	16.0	19.1*	16.0	16.0	16.0	14.8	14.8
##	301	302	303	304	305	306	307	308	309	310	311	312
##	14.8	16.1	19.1*	16.1	18.4	18.4	18.4	18.4	18.4	18.4	18.4	18.4
##	313	314	315	316	317	318	319	320	321	322	323	324
##	19.1*	19.1*	18.4	18.4	18.4	18.4	18.4	18.4	19.6	19.6	19.6	19.6

```
## 325 326 327 328 329 330 331 332 333 334 335 336
## 19.6 19.6 19.1* 19.6 16.9 16.9 16.9 19.1* 16.9 20.2 20.2 20.2
## 337 338 339 340 341 342 343 344 345 346 347 348
## 20.2 20.2 20.2 20.2 20.2 15.5 15.9 17.6 17.6 18.8 18.8 17.9
## 349 350 351 352 353 354 355 356 357 358 359 360
## 17.0 19.7 19.7 18.3 18.3 17.0 22.0 22.0 20.2 20.2 20.2 20.2
## 361 362 363 364 365 366 367 368 369 370 371 372
## 20.2 20.2 20.2 19.1* 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2
## 373 374 375 376 377 378 379 380 381 382 383 384
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2
## 385 386 387 388 389 390 391 392 393 394 395 396
## 20.2 20.2 20.2 20.2 20.2 20.2 19.1* 19.1* 20.2 20.2 20.2 20.2
## 397 398 399 400 401 402 403 404 405 406 407 408
## 20.2 20.2 20.2 20.2 20.2 20.2 19.1* 20.2 20.2 20.2 20.2 20.2
## 409 410 411 412 413 414 415 416 417 418 419 420
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 19.1* 20.2 20.2 20.2
## 421 422 423 424 425 426 427 428 429 430 431 432
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2
## 433 434 435 436 437 438 439 440 441 442 443 444
## 20.2 20.2 20.2 20.2 19.1* 20.2 20.2 20.2 20.2 19.1* 20.2 20.2
## 445 446 447 448 449 450 451 452 453 454 455 456
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2
## 457 458 459 460 461 462 463 464 465 466 467 468
## 20.2 19.1* 20.2 20.2 20.2 20.2 20.2 20.2 19.1* 20.2 20.2 20.2
## 469 470 471 472 473 474 475 476 477 478 479 480
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2
## 481 482 483 484 485 486 487 488 489 490 491 492
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.1 20.1 20.1 20.1
## 493 494 495 496 497 498 499 500 501 502 503 504
## 20.1 19.2 19.2 19.2 19.2 19.2 19.2 19.1* 19.2 21.0 21.0 21.0
## 505 506
## 21.0 21.0
```

```
impute(BostonHousing$ptratio, 20) # replace specific number
```

```
## 1 2 3 4 5 6 7 8 9 10 11 12
## 15.3 17.8 17.8 18.7 18.7 18.7 15.2 15.2 15.2 15.2 15.2 15.2
## 13 14 15 16 17 18 19 20 21 22 23 24
## 15.2 21.0 21.0 21.0 21.0 21.0 21.0 21.0 21.0 21.0 21.0 21.0
## 25 26 27 28 29 30 31 32 33 34 35 36
## 21.0 21.0 21.0 21.0 21.0 21.0 21.0 21.0 21.0 21.0 21.0 19.2
## 37 38 39 40 41 42 43 44 45 46 47 48
## 19.2 19.2 19.2 18.3 18.3 17.9 20.0* 17.9 17.9 17.9 17.9 17.9
## 49 50 51 52 53 54 55 56 57 58 59 60
## 17.9 17.9 16.8 16.8 16.8 16.8 21.1 17.9 17.3 15.1 19.7 19.7
## 61 62 63 64 65 66 67 68 69 70 71 72
## 20.0* 19.7 19.7 19.7 18.6 16.1 16.1 18.9 18.9 18.9 19.2 19.2
## 73 74 75 76 77 78 79 80 81 82 83 84
## 19.2 19.2 18.7 18.7 18.7 18.7 18.7 18.7 19.0 19.0 19.0 19.0
## 85 86 87 88 89 90 91 92 93 94 95 96
## 18.5 18.5 18.5 18.5 17.8 17.8 17.8 17.8 18.2 18.2 18.2 18.0
## 97 98 99 100 101 102 103 104 105 106 107 108
## 18.0 18.0 20.0* 18.0 20.9 20.9 20.0* 20.0* 20.9 20.9 20.9 20.9
## 109 110 111 112 113 114 115 116 117 118 119 120
```

##	20.9	20.9	20.9	17.8	20.0*	17.8	17.8	17.8	20.0*	20.0*	17.8	17.8
##	121	122	123	124	125	126	127	128	129	130	131	132
##	19.1	19.1	19.1	19.1	20.0*	19.1	19.1	21.2	21.2	21.2	21.2	21.2
##	133	134	135	136	137	138	139	140	141	142	143	144
##	21.2	21.2	21.2	20.0*	21.2	21.2	21.2	21.2	21.2	21.2	14.7	14.7
##	145	146	147	148	149	150	151	152	153	154	155	156
##	14.7	14.7	14.7	14.7	14.7	14.7	14.7	14.7	20.0*	14.7	14.7	20.0*
##	157	158	159	160	161	162	163	164	165	166	167	168
##	14.7	14.7	14.7	14.7	14.7	14.7	14.7	20.0*	14.7	14.7	14.7	20.0*
##	169	170	171	172	173	174	175	176	177	178	179	180
##	14.7	14.7	14.7	14.7	20.0*	16.6	16.6	16.6	16.6	16.6	16.6	17.8
##	181	182	183	184	185	186	187	188	189	190	191	192
##	17.8	17.8	17.8	17.8	17.8	17.8	17.8	15.2	15.2	15.2	15.2	15.2
##	193	194	195	196	197	198	199	200	201	202	203	204
##	15.2	15.6	15.6	14.4	20.0*	12.6	12.6	17.0	17.0	14.7	14.7	14.7
##	205	206	207	208	209	210	211	212	213	214	215	216
##	14.7	18.6	18.6	18.6	18.6	18.6	18.6	18.6	18.6	20.0*	20.0*	18.6
##	217	218	219	220	221	222	223	224	225	226	227	228
##	16.4	16.4	20.0*	16.4	17.4	17.4	17.4	17.4	17.4	20.0*	17.4	17.4
##	229	230	231	232	233	234	235	236	237	238	239	240
##	17.4	17.4	17.4	17.4	17.4	17.4	17.4	17.4	17.4	17.4	16.6	16.6
##	241	242	243	244	245	246	247	248	249	250	251	252
##	16.6	16.6	16.6	16.6	19.1	19.1	20.0*	19.1	19.1	19.1	19.1	19.1
##	253	254	255	256	257	258	259	260	261	262	263	264
##	19.1	19.1	16.4	16.4	15.9	13.0	13.0	13.0	13.0	13.0	13.0	13.0
##	265	266	267	268	269	270	271	272	273	274	275	276
##	13.0	13.0	13.0	13.0	13.0	18.6	18.6	20.0*	18.6	18.6	17.6	17.6
##	277	278	279	280	281	282	283	284	285	286	287	288
##	17.6	17.6	17.6	14.9	14.9	14.9	14.9	13.6	15.3	15.3	18.2	16.6
##	289	290	291	292	293	294	295	296	297	298	299	300
##	16.6	16.6	20.0*	20.0*	19.2	16.0	20.0*	16.0	16.0	16.0	14.8	14.8
##	301	302	303	304	305	306	307	308	309	310	311	312
##	14.8	16.1	20.0*	16.1	18.4	18.4	18.4	18.4	18.4	18.4	18.4	18.4
##	313	314	315	316	317	318	319	320	321	322	323	324
##	20.0*	20.0*	18.4	18.4	18.4	18.4	18.4	18.4	19.6	19.6	19.6	19.6
##	325	326	327	328	329	330	331	332	333	334	335	336
##	19.6	19.6	20.0*	19.6	16.9	16.9	16.9	20.0*	16.9	20.2	20.2	20.2
##	337	338	339	340	341	342	343	344	345	346	347	348
##	20.2	20.2	20.2	20.2	20.2	15.5	15.9	17.6	17.6	18.8	18.8	17.9
##	349	350	351	352	353	354	355	356	357	358	359	360
##	17.0	19.7	19.7	18.3	18.3	17.0	22.0	22.0	20.2	20.2	20.2	20.2
##	361	362	363	364	365	366	367	368	369	370	371	372
##	20.2	20.2	20.2	20.0*	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2
##	373	374	375	376	377	378	379	380	381	382	383	384
##	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2
##	385	386	387	388	389	390	391	392	393	394	395	396
##	20.2	20.2	20.2	20.2	20.2	20.2	20.0*	20.0*	20.2	20.2	20.2	20.2
##	397	398	399	400	401	402	403	404	405	406	407	408
##	20.2	20.2	20.2	20.2	20.2	20.2	20.0*	20.2	20.2	20.2	20.2	20.2
##	409	410	411	412	413	414	415	416	417	418	419	420
##	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.0*	20.2	20.2	20.2
##	421	422	423	424	425	426	427	428	429	430	431	432
##	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2	20.2
##	433	434	435	436	437	438	439	440	441	442	443	444

```
## 20.2 20.2 20.2 20.2 20.0* 20.2 20.2 20.2 20.2 20.0* 20.2 20.2
## 445 446 447 448 449 450 451 452 453 454 455 456
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2
## 457 458 459 460 461 462 463 464 465 466 467 468
## 20.2 20.0* 20.2 20.2 20.2 20.2 20.2 20.2 20.0* 20.2 20.2 20.2
## 469 470 471 472 473 474 475 476 477 478 479 480
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2
## 481 482 483 484 485 486 487 488 489 490 491 492
## 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.2 20.1 20.1 20.1 20.1
## 493 494 495 496 497 498 499 500 501 502 503 504
## 20.1 19.2 19.2 19.2 19.2 19.2 19.2 20.0* 19.2 21.0 21.0 21.0
## 505 506
## 21.0 21.0
```

*# or if you want to impute manually*

```
## BostonHousing$ptratio[is.na(BostonHousing$ptratio)] <- mean(BostonHousing$ptratio, na.rm = T) # not
```

Calculemos la precisión cuando se imputa con la media:

```
library(DMwR)
```

```
## Loading required package: grid
```

```
actuals <- original$ptratio[is.na(BostonHousing$ptratio)]
predicted <- rep(mean(BostonHousing$ptratio, na.rm=T), length(actuals))
regr.eval(actuals, predicted)
```

```
##          mae          mse          rmse          mape
## 1.62324034 4.19306071 2.04769644 0.09545664
```

## Imputación basada en predicciones:

### kNN Imputación

```
library(DMwR)
knnOutput <- knnImputation(BostonHousing[, !names(BostonHousing) %in% "medv"], k=5) # perform knn imputation
anyNA(knnOutput)
```

```
## [1] FALSE
```

```
#> FALSE
```

Calculemos la precisión

```
actuals <- original$ptratio[is.na(BostonHousing$ptratio)]
predicted <- knnOutput[is.na(BostonHousing$ptratio), "ptratio"]
regr.eval(actuals, predicted)
```

```
##          mae          mse          rmse          mape
## 0.75909106 1.32301168 1.15022245 0.04398371
```

### rpart

Solo una de las variables predictoras tiene NA y puede tratar con variables factor



```
library(rpart)
class_mod <- rpart(rad ~ . - medv, data=BostonHousing[!is.na(BostonHousing$rad), ], method="class", na.action=na.omit)
anova_mod <- rpart(ptratio ~ . - medv, data=BostonHousing[!is.na(BostonHousing$ptratio), ], method="anova", na.action=na.omit)
rad_pred <- predict(class_mod, BostonHousing[is.na(BostonHousing$rad), ])
ptratio_pred <- predict(anova_mod, BostonHousing[is.na(BostonHousing$ptratio), ])
```

Calculemos la precisión para ptratio

```
actuals <- original$ptratio[is.na(BostonHousing$ptratio)]
predicteds <- ptratio_pred
regr.eval(actuals, predicteds)
```

```
##          mae          mse          rmse          mape
## 0.71061673 0.99693845 0.99846805 0.04099908
```

Para rad:

```
actuals <- original$rad[is.na(BostonHousing$rad)]
predicteds <- as.numeric(colnames(rad_pred)[apply(rad_pred, 1, which.max)])
mean(actuals != predicteds)
```

```
## [1] 0.25
```

## MICE (Multivariate Imputation by Chained Equations)

Produce multiples copias completas de los datos, con valores diferentes sólo para los valores perdidos:

```
library(mice)
library(randomForest)
```

```
## randomForest 4.6-12
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
##
```

```
## Attaching package: 'randomForest'
```

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

```
##
```

```
##      combine
```

```
## The following object is masked from 'package:ggplot2':
```

```
##
```

```
##      margin
```

```
miceMod <- mice(BostonHousing[, !names(BostonHousing) %in% "medv"], method="rf") # perform mice imputation
```

```
##
```

```
## iter imp variable
```

```
## 1 1 rad ptratio
```

```
## 1 2 rad ptratio
```

```
## 1 3 rad ptratio
```

```
## 1 4 rad ptratio
```

```
## 1 5 rad ptratio
```

```
## 2 1 rad ptratio
## 2 2 rad ptratio
## 2 3 rad ptratio
## 2 4 rad ptratio
## 2 5 rad ptratio
## 3 1 rad ptratio
## 3 2 rad ptratio
## 3 3 rad ptratio
## 3 4 rad ptratio
## 3 5 rad ptratio
## 4 1 rad ptratio
## 4 2 rad ptratio
## 4 3 rad ptratio
## 4 4 rad ptratio
## 4 5 rad ptratio
## 5 1 rad ptratio
## 5 2 rad ptratio
## 5 3 rad ptratio
## 5 4 rad ptratio
## 5 5 rad ptratio
```

```
miceOutput <- complete(miceMod) # generate the completed data.
anyNA(miceOutput)
```

```
## [1] FALSE
```

Para evaluar la precisión para ptratio:

```
actuals <- original$ptratio[is.na(BostonHousing$ptratio)]
predicted <- miceOutput[is.na(BostonHousing$ptratio), "ptratio"]
regr.eval(actuals, predicted)
```

```
##          mae          mse          rmse          mape
## 0.88250000 3.31225000 1.81995879 0.05378101
```

Y para rad:

```
actuals <- original$rad[is.na(BostonHousing$rad)]
predicted <- miceOutput[is.na(BostonHousing$rad), "rad"]
mean(actuals != predicted)
```

```
## [1] 0.225
```