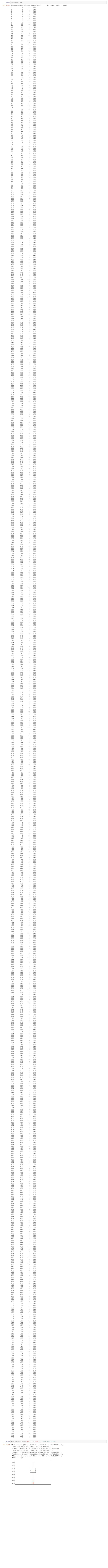
l	<pre>import pandas as pd # importing pandas = &gt; useful for creating dataframes import numpy as np  x1 = [1, 2, 3, 4,5] # list format x2 = [10, 11, 12,100,11]</pre> x1
[5]: [ [6]: [ t[6]: [ 7]: [	<pre>[1, 2, 3, 4, 5]  x3 = list(range(5))  x3  [0, 1, 2, 3, 4]  X = pd.DataFrame(columns = ["X1", "X2", "X3"])</pre>
[7]:	<pre>X</pre>
[8]: ( t[8]:	X1       X2       X3         0       1       10       0         1       2       11       1         2       3       12       2         3       4       100       3
	X["X1"] = x1 # Converting list format into pandas series format X["X2"] = x2 # Converting list format into pandas series format X["X3"] = x3  X
[11]: [ [11]:	0 1 1 2 2 3 3 4 4 5
	<pre>4    5 Name: X1, dtype: int64  X[["X1","X2"]]  X1</pre>
[13]: [ [13]:	2 3 12 3 4 100 4 5 11 X.iloc[0:3,1] 0 10 1 11
[9]:[ t[9]:	1 11 2 12 Name: X2, dtype: int64  X.iloc[:,:]  X1 X2 X3
[15]:   [15]:	X.loc[0:3,["X1","X2"]]  X1 X2  0 1 10  1 2 11  2 3 12  3 4 100
[16]:	<pre>3  4 100  type(X.X1)  pandas.core.series.Series  x = pd.DataFrame(columns=["A", "B", "C"])</pre>
18]: [ 18]:	X
.9]: [ 20]: [	4 5 11 4  x["A"] = pd.Series(list(np.random.randint(1,100,50)))  x  A B C
	0 28 NaN NaN 1 93 NaN NaN 2 98 NaN NaN 3 92 NaN NaN 4 49 NaN NaN 5 89 NaN NaN
	6 13 NaN NaN 7 30 NaN NaN 8 52 NaN NaN 9 57 NaN NaN 10 53 NaN NaN 11 55 NaN NaN
	<ul> <li>12 14 NaN NaN</li> <li>13 32 NaN NaN</li> <li>14 47 NaN NaN</li> <li>15 55 NaN NaN</li> <li>16 54 NaN NaN</li> <li>17 82 NaN NaN</li> </ul>
	<ul> <li>18 24 NaN NaN</li> <li>19 25 NaN NaN</li> <li>20 27 NaN NaN</li> <li>21 15 NaN NaN</li> <li>22 29 NaN NaN</li> <li>23 62 NaN NaN</li> </ul>
	<ul> <li>24 7 NaN NaN</li> <li>25 65 NaN NaN</li> <li>26 37 NaN NaN</li> <li>27 61 NaN NaN</li> <li>28 93 NaN NaN</li> <li>29 53 NaN NaN</li> </ul>
	30 2 NaN NaN 31 71 NaN NaN 32 88 NaN NaN 33 89 NaN NaN 34 81 NaN NaN 35 84 NaN NaN 36 43 NaN NaN
	37 40 NaN NaN 38 9 NaN NaN 39 46 NaN NaN 40 41 NaN NaN 41 42 NaN NaN 42 64 NaN NaN
	<ul> <li>43 40 NaN NaN</li> <li>44 41 NaN NaN</li> <li>45 23 NaN NaN</li> <li>46 17 NaN NaN</li> <li>47 54 NaN NaN</li> <li>48 54 NaN NaN</li> </ul>
l	49 60 NaN NaN  x["B"] = pd.Series(list(np.random.choice([0,1], size = 50)))  x  A B C
	0 28 1 NaN 1 93 0 NaN 2 98 1 NaN 3 92 1 NaN 4 49 1 NaN 5 89 0 NaN
	6 13 1 NaN 7 30 0 NaN 8 52 1 NaN 9 57 0 NaN 10 53 0 NaN 11 55 1 NaN
	12 14 0 NaN 13 32 1 NaN 14 47 0 NaN 15 55 0 NaN 16 54 0 NaN 17 82 0 NaN 18 24 0 NaN
	19
	25 65 1 NaN 26 37 0 NaN 27 61 1 NaN 28 93 1 NaN 29 53 0 NaN 30 2 1 NaN
	31 71 1 NaN 32 88 0 NaN 33 89 1 NaN 34 81 1 NaN 35 84 1 NaN 36 43 0 NaN
	37 40 1 NaN 38 9 1 NaN 39 46 1 NaN 40 41 0 NaN 41 42 1 NaN 42 64 1 NaN
	43 40 0 NaN 44 41 0 NaN 45 23 1 NaN 46 17 0 NaN 47 54 0 NaN 48 54 1 NaN 49 60 1 NaN
]:[	x["C"] = 15
	0       28       1       15         1       93       0       15         2       98       1       15         3       92       1       15         4       49       1       15         5       89       0       15         6       13       1       15
	7       30       0       15         8       52       1       15         9       57       0       15         10       53       0       15         11       55       1       15         12       14       0       15
	13       32       1       15         14       47       0       15         15       55       0       15         16       54       0       15         17       82       0       15         18       24       0       15
	19 25 0 15 20 27 0 15 21 15 1 15 22 29 1 15 23 62 1 15 24 7 1 15 25 65 1 15
	38       9       1       15         39       46       1       15         40       41       0       15         41       42       1       15         42       64       1       15         43       40       0       15
	44       41       0       15         45       23       1       15         46       17       0       15         47       54       0       15         48       54       1       15         49       60       1       15
	mba  NameError Traceback (most recent call last) <ipython-input-10-60f45ea320a1> in <module>&gt; 1 mba  NameError: name 'mba' is not defined</module></ipython-input-10-60f45ea320a1>
8]:	<pre>import pandas as pd import numpy as np import matplotlib.pyplot as plt  mba = pd.read_csv("file:///C:/Users/Mukesh/Downloads/mba.csv") type(mba)</pre>
	<pre>uoting, doublequote, escapechar, comment, encoding, dialect, error_bad_lines, warn_bad_lines, delim_w hitespace, low_memory, memory_map, float_precision) 683 684 -&gt; 685 return_read(filepath_or_buffer, kwds) 686 687 parser_fname = name  ~\Anaconda3\lib\site-packages\pandas\io\parsers.py in read(filepath_or_buffer, kwds) 438 # See https://github.com/python/mypy/issues/1297 439 fp_or_buf,compression, should_close = get_filepath_or_buffer( -&gt; 440 filepath_or_buffer, encoding, compression  441 442 kwds["compression"] = compression  ~\Anaconda3\lib\site-packages\pandas\io\common.py in get_filepath_or_buffer(filepath_or_buffer, encoding, compression, mode) 194 195 ifis_url(filepath_or_buffer): -&gt; 196 req = urlopen(filepath_or_buffer) 197 content_encoding = req.headers.get("Content-Encoding", None) 198 if content_encoding == "gaip":  ~\Anaconda3\lib\urllib\request.py in urlopen(url, data, timeout, cafile, capath, cadefault, context) 220 else: 221</pre>
	hitespace, low_memory, memory_map, float_precision) 683 684 685 686 687 parser_fname_ = name  "Anaconda3\lib\site-packages\pandas\io\parsers.py in road(filepath_or_buffer, kwds) 488 48
	hitespace, low_memory, map, float_precision)  883  884  886  887  Parser_fname = name  -\nanconda3\lib\nite-packages\pandes\is\parser.py in _rewi(filepath_or_buffer, kwds)  887  -\nanconda3\lib\nite-packages\pandes\is\parser.py in _rewi(filepath_or_buffer, kwds)  888  889    a
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<pre>In [32]: [ Out[32]:</pre>	0 1 2 3	datasrno  1 2 3 4	21 107 57	720 640
	3 4 5 6 7 8	4 5 6 7 8 9	208 136 70 103	710 660
	9 10 11 12 13	10 11 12 13 14	22 69 41 72 69	730 700 740 650 720
	14 15 16 17 18	15 16 17 18 19 20	21 19 86 231	700 760 710 720 690 770
	20 21 22 23 24 25	21 22 23 24 25 26	21 44 23 20	650 710
	26 27 28 29 30	27 28 29 30 31	46 33 130 57	710 650 690 690 750
	31 32 33 34 35 36	32 33 34 35 36 37	55 42 34 55	770 690 710
	37 38 39 40 41 42	38 39 40 41 42 43	45 38 44 83	740 720
	43 44 45 46 47	44 45 46 47 48	45 89 77 91 61	710 690 700 730 690
	48 49 50 51 52	49 50 51 52 53	69 59 32 74	710 720 710 740
	53 54 55 56 57 58	54 55 56 57 58 59	33 33 34 110	730 730 710
	59 60 61 62 63	60 61 62 63 64	44 46 58 45 34	690 720 690 710 720
	64 65 66 67 68	65 66 67 68 69	45 33 82 72	720 740 710 740
	69 70 71 72 73	70 71 72 73 74	43 54 92 90	740 740 730 720
	74 75 76 77 78 79	75 76 77 78 79 80	38 25 57 31	740
	80 81 82 83 84	81 82 83 84 85	55 68 79 71	710 710 730
	85 86 87 88 89 90	86 87 88 89 90	32 45 69 65	710 710 720 680
	91 92 93 94 95	92 93 94 95 96	45 92 57 69	710
	96 97 98 99	97 98 99 100 101	59 44 44 46	700 690 770 770 720
	101 102 103 104 105 106	102 103 104 105 106 107	46 82 53 33	730 710 720 690 740 690
	107 108 109 110 111	108 109 110 111 112	37 55 52 68	690 750 720 670 690
	112 113 114 115 116 117	113 114 115 116 117 118	47 56 45	710 690 750 710 680 690
	117 118 119 120 121 122	119 120 121 122 123	34 55 43	730 750 700 720 680
	123 124 125 126 127	124 125 126 127 128	47 56	690 700 740 710 730
	128 129 130 131 132	129 130 131 132 133	46 58 34	720 720 740 680 670
	133 134 135 136 137	134 135 136 137 138	68 34 46	730 680 710 740 760
	138 139 140 141 142 143	139 140 141 142 143 144	56 33 33 44	710 710 710 730 680 690
	143 144 145 146 147 148	144 145 146 147 148 149	69 82 64 126	720 730
	<ul><li>149</li><li>150</li><li>151</li><li>152</li><li>153</li></ul>	150 151 152 153 154	80 45 45 68	720 710 700 720 720
	154 155 156 157 158	159	130 80 130 34	690 710 730 710 740
	159 160 161 162 163 164	160 161 162 163 164	48 45 40 58	690 630 730 720 710
	164 165 166 167 168 169	165 166 167 168 169 170	40 79 34 56 68	690 690 730 710 720 710
	170 171 172 173 174	171 172 173 174 175	32 57 53 57 45	740 710 650 760 680
	174 175 176 177 178 179	176 177 178 179 180	46 77 44 68 45	700 710 690 720 740
	180 181 182 183 184	181 182 183 184 185	54 58 59 38 94	710 710 730 660 710
	185 186 187 188 189	186 187 188 189 190	46 96 44 50 91	710 710 740 690 610
	190 191 192 193 194 195	191 192 193 194 195	117 58 32 57	700 690 760 730 700
			57 32 33 41 70	720 710 750 690 750
	<ul><li>200</li><li>201</li><li>202</li><li>203</li><li>204</li></ul>	201 202 203 204 205	45 68 90 82 33	660 740 710 730 750
	204 205 206 207 208 209 210	205 206 207 208 209 210 211	46 45 45 79 109	710 710 710 680 720
	<ul><li>210</li><li>211</li><li>212</li><li>213</li><li>214</li><li>215</li></ul>	211 212 213 214 215 216	44 33 70 44	700 740 690 710
	216 217 218 219 220	217 218 219 220 221	85 57 66 44	720 710 710 700
	<ul><li>221</li><li>222</li><li>223</li><li>224</li><li>225</li></ul>	222 223 224 225 226	33 89 55 128	710 710 710
	<ul><li>226</li><li>227</li><li>228</li><li>229</li><li>230</li><li>231</li></ul>	229 230 231	33 35 33 58	710 710 720 710 660 650
	<ul><li>231</li><li>232</li><li>233</li><li>234</li><li>235</li><li>236</li></ul>	232 233 234 235 236 237	57 45 63 81	
	<ul><li>237</li><li>238</li><li>239</li><li>240</li><li>241</li></ul>	238 239 240 241 242	45 86 45 69	690 760 740 750 710
	<ul><li>242</li><li>243</li><li>244</li><li>245</li><li>246</li></ul>	243 244 245 246 247	70 33 43	710
	<ul><li>247</li><li>248</li><li>249</li><li>250</li><li>251</li></ul>	248 249 250 251 252	57 70 93	690 700 700
	<ul><li>252</li><li>253</li><li>254</li><li>255</li><li>256</li></ul>	253 254 255 256 257	80 33 32 55	720 760 710 690
	<ul><li>257</li><li>258</li><li>259</li><li>260</li><li>261</li></ul>	258 259 260 261 262	66 65 42 71	720 720
	<ul><li>262</li><li>263</li><li>264</li><li>265</li><li>266</li></ul>	263 264 265 266 267	30 45 91 46	730 730 720 740
	<ul><li>267</li><li>268</li><li>269</li><li>270</li><li>271</li></ul>	268 269 270 271 272	74 45 112 46	710 710 720 700
	<ul><li>272</li><li>273</li><li>274</li><li>275</li><li>276</li></ul>	273 274 275 276 277	32 69 44 69	770 760 740 700
	<ul><li>277</li><li>278</li><li>279</li><li>280</li><li>281</li></ul>	278 279 280 281 282	81 82 44 34	690 700
	282 283 284 285 286	283 284 285 286 287	34 43 69 34	730 720 720 710 720
	287 288 289 290 291	288 289 290 291 292	45 50 57 50	680
	<ul><li>292</li><li>293</li><li>294</li><li>295</li><li>296</li></ul>	293 294 295 296 297	92 131 45 64	
	<ul><li>297</li><li>298</li><li>299</li><li>300</li><li>301</li></ul>	298 299 300 301 302	32 100	
	302 303 304 305 306 307	303 304 305 306 307 308	45 68 28 57	710 710 700 680 700 680
	308 309 310 311 312	309 310 311 312 313	58 46 142 47	650 690 710 660
	<ul><li>313</li><li>314</li><li>315</li><li>316</li><li>317</li></ul>	314 315 316 317 318	55 32 51	710 690 700 720 730
	318 319 320 321 322	319 320 321 322 323	53 57 53 56 88	720 720 710 710 690
	323 324 325 326 327	324 325 326 327 328	62 69 91 93 106	710 710 760 700 760
	328 329 330 331 332	329 330 331 332 333	68 56 57 56 70	760 740 710 730 700
	333 334 335 336 337 338	334 335 336 337 338 339	70 40 46 45 52	730 710 700 710 620
	338 339 340 341 342	339 340 341 342 343	80 69 56 82 58	660 710 730 680 740
	343 344 345 346 347 348	344 345 346 347 348 349	68 39 92 34 42	710 710 710 750
			69 33 268 70 54	710 740 730
	354 355 356 357 358	355 356 357 358 359	42 44 57 81 103	760 730 720 740 740
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	386 387 388 389 390	387 388 389 390 391	37 33 81 131 43	720 740 750 730 680
	391 392 393 394 395 396	392 393 394 395 396 397	69 72 34 58 69	680
	397 398 399 400 401	398 399 400 401 402	116 77 45 57 56	720 700 680 700 730
	402 403 404 405 406	403 404 405 406 407	57 109 44 37 45	710 620 750 720 740
	407 408 409 410 411	408 409 410 411 412	46 68 78 63 45	700 720 720 730 740
	412 413 414 415 416	413 414 415 416 417	67 33 69 45 176	690 760 730 720 710
	417 418 419 420 421 422	418 419 420 421 422 423	82 55 32 45	770 720 710 720 720
			69 32 93 131 49	770
			39 37 111 66 45	690 680 710 750 730
	432 433 434 435 436	433 434 435 436 437	44 35 67 33 57	730 720 710 710 710
	437 438 439 440 441 442	438 439 440 441 442 443	52 49 29 31	730 760 700 720 720
			34 46 46 122	690 700 700 720
			29 70 43 116	650
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howbox=None, showfliers=None, boxprops=None, labels=None, flierprops=None, medianprops=None, mes=None, capprops=None, whiskerprops=None, manage_ticks=True, autorange=False, zorder=None, *, cone)  Make a box and whisker plot.  Make a box and whisker plot for each column of ``x`` or each vector in sequence ``x``. The box extends from the lower to
upper quartile values of the data, with a line at the median.  The whiskers extend from the box to show the range of the data. Flier points are those past the end of the whiskers.  Parameters x: Array or a sequence of vectors. The input data.  notch: bool, optional (False)
If `True`, will produce a notched box plot. Otherwise, a rectangular boxplot is produced. The notches represent the confidence interval (CI) around the median. See the entry for the ``bootstrap`` parameter for information regarding how the locations of the notches are computed.  note::  In cases where the values of the CI are less than the lower quartile or greater than the upper quartile, the
notches will extend beyond the box, giving it a distinctive "flipped" appearance. This is expected behavior and consistent with other statistical visualization packages.  sym: str, optional The default symbol for flier points. Enter an empty string ('') if you don't want to show fliers. If `None`, then the fliers default to 'b+' If you want more control use the
<pre>flierprops kwarg.  vert : bool, optional (True)    If `True` (default), makes the boxes vertical. If `False`,    everything is drawn horizontally.  whis : float, sequence, or string (default = 1.5)    As a float, determines the reach of the whiskers to the beyond the    first and third quartiles. In other words, where IQR is the    interquartile range (`Q3-Q1`), the upper whisker will extend to</pre>
last datum less than `Q3 + whis*IQR`). Similarly, the lower whisker will extend to the first datum greater than `Q1 - whis*IQR`. Beyond the whiskers, data are considered outliers and are plotted as individual points. Set this to an unreasonably high value to force the whiskers to show the min and max values. Alternatively, set this to an ascending sequence of percentile (e.g., [5, 95]) to set the whiskers at specific percentiles of the data. Finally, ``whis`` can be the string ``'range'`` to force the whiskers to the min and max of the data.
bootstrap: int, optional  Specifies whether to bootstrap the confidence intervals  around the median for notched boxplots. If ``bootstrap`` is  None, no bootstrapping is performed, and notches are  calculated using a Gaussian-based asymptotic approximation  (see McGill, R., Tukey, J.W., and Larsen, W.A., 1978, and  Kendall and Stuart, 1967). Otherwise, bootstrap specifies  the number of times to bootstrap the median to determine its  95% confidence intervals. Values between 1000 and 10000 are
recommended.  usermedians : array-like, optional  An array or sequence whose first dimension (or length) is compatible with ``x``. This overrides the medians computed by matplotlib for each element of ``usermedians`` that is not `None`. When an element of ``usermedians`` is None, the median will be computed by matplotlib as normal.  conf_intervals : array-like, optional
Array or sequence whose first dimension (or length) is compatible with ``x`` and whose second dimension is 2. When the an element of ``conf_intervals`` is not None, the notch locations computed by matplotlib are overridden (provided ``notch`` is `True`). When an element of ``conf_intervals`` is `None`, the notches are computed by the method specified by the other kwargs (e.g., ``bootstrap``).  positions: array-like, optional
Sets the positions of the boxes. The ticks and limits are automatically set to match the positions. Defaults to 'range(1, N+1)' where N is the number of boxes to be drawn.  widths: scalar or array-like  Sets the width of each box either with a scalar or a sequence. The default is 0.5, or ``0.15*(distance between extreme positions)``, if that is smaller.  patch_artist: bool, optional (False)
<pre>If `False` produces boxes with the Line2D artist. Otherwise,   boxes and drawn with Patch artists.  labels : sequence, optional   Labels for each dataset. Length must be compatible with   dimensions of ``x``.  manage_ticks : bool, optional (True)   If True, the tick locations and labels will be adjusted to match   the boxplot positions.</pre>
autorange: bool, optional (False)  When `True` and the data are distributed such that the 25th and 75th percentiles are equal, ``whis`` is set to ``'range'`` such that the whisker ends are at the minimum and maximum of the data.  meanline: bool, optional (False) If `True` (and ``showmeans`` is `True`), will try to render the mean as a line spanning the full width of the box
according to ``meanprops`` (see below). Not recommended if   ``shownotches`` is also True. Otherwise, means will be shown   as points.  zorder: scalar, optional (None)   Sets the zorder of the boxplot.  Other Parameters
showcaps : bool, optional (True) Show the caps on the ends of whiskers. showbox : bool, optional (True) Show the central box. showfliers : bool, optional (True) Show the outliers beyond the caps. showmeans : bool, optional (False) Show the arithmetic means. capprops : dict, optional (None) Specifies the style of the caps.
boxprops : dict, optional (None)  Specifies the style of the box.  whiskerprops : dict, optional (None)  Specifies the style of the whiskers.  flierprops : dict, optional (None)  Specifies the style of the fliers.  medianprops : dict, optional (None)  Specifies the style of the median.  meanprops : dict, optional (None)  Specifies the style of the mean.
Returns result : dict A dictionary mapping each component of the boxplot to a list of the :class:`matplotlib.lines.Line2D` instances created. That dictionary has the following keys (assuming vertical boxplots):
<ul> <li>'boxes': the main body of the boxplot showing the quartiles and the median's confidence intervals if enabled.</li> <li>'medians': horizontal lines at the median of each box.</li> <li>'whiskers': the vertical lines extending to the most extreme, non-outlier data points.</li> <li>'caps': the horizontal lines at the ends of the</li> </ul>
<pre>whiskers.  - ``fliers``: points representing data that extend beyond     the whiskers (fliers).  - ``means``: points or lines representing the means.</pre> Notes
<pre> note::     In addition to the above described arguments, this function can take a     **data** keyword argument. If such a **data** argument is given, the     following arguments are replaced by **data[<arg>]**:     * All positional and all keyword arguments.  Objects passed as **data** must support item access (``data[<arg>]``) and</arg></arg></pre>
membership test (`` <arg> in data``).</arg>

In [15]: help(plt.boxplot)

