

# KNN Case Study

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02/06/2021

```
#import dataset
```

```
dataset <- read.csv("Social_Network_Ads.csv")  
dataset = dataset[3:5]  
dataset
```

##	Age	EstimatedSalary	Purchased
## 1	19	19000	0
## 2	35	20000	0
## 3	26	43000	0
## 4	27	57000	0
## 5	19	76000	0
## 6	27	58000	0
## 7	27	84000	0
## 8	32	150000	1
## 9	25	33000	0
## 10	35	65000	0
## 11	26	80000	0
## 12	26	52000	0
## 13	20	86000	0
## 14	32	18000	0
## 15	18	82000	0
## 16	29	80000	0
## 17	47	25000	1
## 18	45	26000	1
## 19	46	28000	1
## 20	48	29000	1
## 21	45	22000	1
## 22	47	49000	1
## 23	48	41000	1
## 24	45	22000	1
## 25	46	23000	1
## 26	47	20000	1
## 27	49	28000	1
## 28	47	30000	1
## 29	29	43000	0
## 30	31	18000	0
## 31	31	74000	0
## 32	27	137000	1
## 33	21	16000	0
## 34	28	44000	0

## 35	27	90000	0
## 36	35	27000	0
## 37	33	28000	0
## 38	30	49000	0
## 39	26	72000	0
## 40	27	31000	0
## 41	27	17000	0
## 42	33	51000	0
## 43	35	108000	0
## 44	30	15000	0
## 45	28	84000	0
## 46	23	20000	0
## 47	25	79000	0
## 48	27	54000	0
## 49	30	135000	1
## 50	31	89000	0
## 51	24	32000	0
## 52	18	44000	0
## 53	29	83000	0
## 54	35	23000	0
## 55	27	58000	0
## 56	24	55000	0
## 57	23	48000	0
## 58	28	79000	0
## 59	22	18000	0
## 60	32	117000	0
## 61	27	20000	0
## 62	25	87000	0
## 63	23	66000	0
## 64	32	120000	1
## 65	59	83000	0
## 66	24	58000	0
## 67	24	19000	0
## 68	23	82000	0
## 69	22	63000	0
## 70	31	68000	0
## 71	25	80000	0
## 72	24	27000	0
## 73	20	23000	0
## 74	33	113000	0
## 75	32	18000	0
## 76	34	112000	1
## 77	18	52000	0
## 78	22	27000	0
## 79	28	87000	0
## 80	26	17000	0
## 81	30	80000	0
## 82	39	42000	0
## 83	20	49000	0
## 84	35	88000	0
## 85	30	62000	0
## 86	31	118000	1
## 87	24	55000	0
## 88	28	85000	0

## 89	26	81000	0
## 90	35	50000	0
## 91	22	81000	0
## 92	30	116000	0
## 93	26	15000	0
## 94	29	28000	0
## 95	29	83000	0
## 96	35	44000	0
## 97	35	25000	0
## 98	28	123000	1
## 99	35	73000	0
## 100	28	37000	0
## 101	27	88000	0
## 102	28	59000	0
## 103	32	86000	0
## 104	33	149000	1
## 105	19	21000	0
## 106	21	72000	0
## 107	26	35000	0
## 108	27	89000	0
## 109	26	86000	0
## 110	38	80000	0
## 111	39	71000	0
## 112	37	71000	0
## 113	38	61000	0
## 114	37	55000	0
## 115	42	80000	0
## 116	40	57000	0
## 117	35	75000	0
## 118	36	52000	0
## 119	40	59000	0
## 120	41	59000	0
## 121	36	75000	0
## 122	37	72000	0
## 123	40	75000	0
## 124	35	53000	0
## 125	41	51000	0
## 126	39	61000	0
## 127	42	65000	0
## 128	26	32000	0
## 129	30	17000	0
## 130	26	84000	0
## 131	31	58000	0
## 132	33	31000	0
## 133	30	87000	0
## 134	21	68000	0
## 135	28	55000	0
## 136	23	63000	0
## 137	20	82000	0
## 138	30	107000	1
## 139	28	59000	0
## 140	19	25000	0
## 141	19	85000	0
## 142	18	68000	0

##	143	35	59000	0
##	144	30	89000	0
##	145	34	25000	0
##	146	24	89000	0
##	147	27	96000	1
##	148	41	30000	0
##	149	29	61000	0
##	150	20	74000	0
##	151	26	15000	0
##	152	41	45000	0
##	153	31	76000	0
##	154	36	50000	0
##	155	40	47000	0
##	156	31	15000	0
##	157	46	59000	0
##	158	29	75000	0
##	159	26	30000	0
##	160	32	135000	1
##	161	32	100000	1
##	162	25	90000	0
##	163	37	33000	0
##	164	35	38000	0
##	165	33	69000	0
##	166	18	86000	0
##	167	22	55000	0
##	168	35	71000	0
##	169	29	148000	1
##	170	29	47000	0
##	171	21	88000	0
##	172	34	115000	0
##	173	26	118000	0
##	174	34	43000	0
##	175	34	72000	0
##	176	23	28000	0
##	177	35	47000	0
##	178	25	22000	0
##	179	24	23000	0
##	180	31	34000	0
##	181	26	16000	0
##	182	31	71000	0
##	183	32	117000	1
##	184	33	43000	0
##	185	33	60000	0
##	186	31	66000	0
##	187	20	82000	0
##	188	33	41000	0
##	189	35	72000	0
##	190	28	32000	0
##	191	24	84000	0
##	192	19	26000	0
##	193	29	43000	0
##	194	19	70000	0
##	195	28	89000	0
##	196	34	43000	0

##	197	30	79000	0
##	198	20	36000	0
##	199	26	80000	0
##	200	35	22000	0
##	201	35	39000	0
##	202	49	74000	0
##	203	39	134000	1
##	204	41	71000	0
##	205	58	101000	1
##	206	47	47000	0
##	207	55	130000	1
##	208	52	114000	0
##	209	40	142000	1
##	210	46	22000	0
##	211	48	96000	1
##	212	52	150000	1
##	213	59	42000	0
##	214	35	58000	0
##	215	47	43000	0
##	216	60	108000	1
##	217	49	65000	0
##	218	40	78000	0
##	219	46	96000	0
##	220	59	143000	1
##	221	41	80000	0
##	222	35	91000	1
##	223	37	144000	1
##	224	60	102000	1
##	225	35	60000	0
##	226	37	53000	0
##	227	36	126000	1
##	228	56	133000	1
##	229	40	72000	0
##	230	42	80000	1
##	231	35	147000	1
##	232	39	42000	0
##	233	40	107000	1
##	234	49	86000	1
##	235	38	112000	0
##	236	46	79000	1
##	237	40	57000	0
##	238	37	80000	0
##	239	46	82000	0
##	240	53	143000	1
##	241	42	149000	1
##	242	38	59000	0
##	243	50	88000	1
##	244	56	104000	1
##	245	41	72000	0
##	246	51	146000	1
##	247	35	50000	0
##	248	57	122000	1
##	249	41	52000	0
##	250	35	97000	1

##	251	44	39000	0
##	252	37	52000	0
##	253	48	134000	1
##	254	37	146000	1
##	255	50	44000	0
##	256	52	90000	1
##	257	41	72000	0
##	258	40	57000	0
##	259	58	95000	1
##	260	45	131000	1
##	261	35	77000	0
##	262	36	144000	1
##	263	55	125000	1
##	264	35	72000	0
##	265	48	90000	1
##	266	42	108000	1
##	267	40	75000	0
##	268	37	74000	0
##	269	47	144000	1
##	270	40	61000	0
##	271	43	133000	0
##	272	59	76000	1
##	273	60	42000	1
##	274	39	106000	1
##	275	57	26000	1
##	276	57	74000	1
##	277	38	71000	0
##	278	49	88000	1
##	279	52	38000	1
##	280	50	36000	1
##	281	59	88000	1
##	282	35	61000	0
##	283	37	70000	1
##	284	52	21000	1
##	285	48	141000	0
##	286	37	93000	1
##	287	37	62000	0
##	288	48	138000	1
##	289	41	79000	0
##	290	37	78000	1
##	291	39	134000	1
##	292	49	89000	1
##	293	55	39000	1
##	294	37	77000	0
##	295	35	57000	0
##	296	36	63000	0
##	297	42	73000	1
##	298	43	112000	1
##	299	45	79000	0
##	300	46	117000	1
##	301	58	38000	1
##	302	48	74000	1
##	303	37	137000	1
##	304	37	79000	1

##	305	40	60000	0
##	306	42	54000	0
##	307	51	134000	0
##	308	47	113000	1
##	309	36	125000	1
##	310	38	50000	0
##	311	42	70000	0
##	312	39	96000	1
##	313	38	50000	0
##	314	49	141000	1
##	315	39	79000	0
##	316	39	75000	1
##	317	54	104000	1
##	318	35	55000	0
##	319	45	32000	1
##	320	36	60000	0
##	321	52	138000	1
##	322	53	82000	1
##	323	41	52000	0
##	324	48	30000	1
##	325	48	131000	1
##	326	41	60000	0
##	327	41	72000	0
##	328	42	75000	0
##	329	36	118000	1
##	330	47	107000	1
##	331	38	51000	0
##	332	48	119000	1
##	333	42	65000	0
##	334	40	65000	0
##	335	57	60000	1
##	336	36	54000	0
##	337	58	144000	1
##	338	35	79000	0
##	339	38	55000	0
##	340	39	122000	1
##	341	53	104000	1
##	342	35	75000	0
##	343	38	65000	0
##	344	47	51000	1
##	345	47	105000	1
##	346	41	63000	0
##	347	53	72000	1
##	348	54	108000	1
##	349	39	77000	0
##	350	38	61000	0
##	351	38	113000	1
##	352	37	75000	0
##	353	42	90000	1
##	354	37	57000	0
##	355	36	99000	1
##	356	60	34000	1
##	357	54	70000	1
##	358	41	72000	0

##	359	40	71000	1
##	360	42	54000	0
##	361	43	129000	1
##	362	53	34000	1
##	363	47	50000	1
##	364	42	79000	0
##	365	42	104000	1
##	366	59	29000	1
##	367	58	47000	1
##	368	46	88000	1
##	369	38	71000	0
##	370	54	26000	1
##	371	60	46000	1
##	372	60	83000	1
##	373	39	73000	0
##	374	59	130000	1
##	375	37	80000	0
##	376	46	32000	1
##	377	46	74000	0
##	378	42	53000	0
##	379	41	87000	1
##	380	58	23000	1
##	381	42	64000	0
##	382	48	33000	1
##	383	44	139000	1
##	384	49	28000	1
##	385	57	33000	1
##	386	56	60000	1
##	387	49	39000	1
##	388	39	71000	0
##	389	47	34000	1
##	390	48	35000	1
##	391	48	33000	1
##	392	47	23000	1
##	393	45	45000	1
##	394	60	42000	1
##	395	39	59000	0
##	396	46	41000	1
##	397	51	23000	1
##	398	50	20000	1
##	399	36	33000	0
##	400	49	36000	1

```

#encoding the features
dataset$Purchased = factor(dataset$Purchased, levels=c(0,1))
#split the dataset
library(caTools)
set.seed(1)
split = sample.split(dataset$Purchased, SplitRatio = 0.75)
training_set = subset(dataset, split ==TRUE)
test_set = subset(dataset, split == FALSE)
training_set

```

```
##      Age EstimatedSalary Purchased
```



## 1	19	19000	0
## 2	35	20000	0
## 3	26	43000	0
## 5	19	76000	0
## 8	32	150000	1
## 9	25	33000	0
## 10	35	65000	0
## 11	26	80000	0
## 12	26	52000	0
## 13	20	86000	0
## 14	32	18000	0
## 15	18	82000	0
## 17	47	25000	1
## 19	46	28000	1
## 20	48	29000	1
## 21	45	22000	1
## 23	48	41000	1
## 24	45	22000	1
## 25	46	23000	1
## 27	49	28000	1
## 28	47	30000	1
## 29	29	43000	0
## 30	31	18000	0
## 32	27	137000	1
## 33	21	16000	0
## 36	35	27000	0
## 37	33	28000	0
## 38	30	49000	0
## 39	26	72000	0
## 40	27	31000	0
## 41	27	17000	0
## 42	33	51000	0
## 44	30	15000	0
## 45	28	84000	0
## 46	23	20000	0
## 47	25	79000	0
## 48	27	54000	0
## 49	30	135000	1
## 51	24	32000	0
## 53	29	83000	0
## 54	35	23000	0
## 55	27	58000	0
## 57	23	48000	0
## 59	22	18000	0
## 60	32	117000	0
## 62	25	87000	0
## 63	23	66000	0
## 64	32	120000	1
## 66	24	58000	0
## 67	24	19000	0
## 69	22	63000	0
## 70	31	68000	0
## 71	25	80000	0
## 72	24	27000	0

## 73	20	23000	0
## 74	33	113000	0
## 75	32	18000	0
## 76	34	112000	1
## 77	18	52000	0
## 79	28	87000	0
## 80	26	17000	0
## 81	30	80000	0
## 82	39	42000	0
## 83	20	49000	0
## 84	35	88000	0
## 86	31	118000	1
## 87	24	55000	0
## 89	26	81000	0
## 91	22	81000	0
## 92	30	116000	0
## 93	26	15000	0
## 96	35	44000	0
## 98	28	123000	1
## 100	28	37000	0
## 101	27	88000	0
## 102	28	59000	0
## 103	32	86000	0
## 104	33	149000	1
## 106	21	72000	0
## 107	26	35000	0
## 108	27	89000	0
## 109	26	86000	0
## 110	38	80000	0
## 111	39	71000	0
## 112	37	71000	0
## 113	38	61000	0
## 117	35	75000	0
## 118	36	52000	0
## 120	41	59000	0
## 121	36	75000	0
## 122	37	72000	0
## 123	40	75000	0
## 125	41	51000	0
## 126	39	61000	0
## 127	42	65000	0
## 128	26	32000	0
## 130	26	84000	0
## 132	33	31000	0
## 133	30	87000	0
## 134	21	68000	0
## 135	28	55000	0
## 136	23	63000	0
## 137	20	82000	0
## 138	30	107000	1
## 139	28	59000	0
## 140	19	25000	0
## 141	19	85000	0
## 143	35	59000	0

##	144	30	89000	0
##	145	34	25000	0
##	147	27	96000	1
##	148	41	30000	0
##	149	29	61000	0
##	150	20	74000	0
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##	152	41	45000	0
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##	155	40	47000	0
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##	161	32	100000	1
##	162	25	90000	0
##	164	35	38000	0
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##	166	18	86000	0
##	167	22	55000	0
##	168	35	71000	0
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##	170	29	47000	0
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##	174	34	43000	0
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##	186	31	66000	0
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##	194	19	70000	0
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##	197	30	79000	0
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##	201	35	39000	0
##	203	39	134000	1
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##	205	58	101000	1
##	207	55	130000	1
##	208	52	114000	0
##	209	40	142000	1
##	211	48	96000	1
##	212	52	150000	1

##	213	59	42000	0
##	214	35	58000	0
##	216	60	108000	1
##	217	49	65000	0
##	219	46	96000	0
##	220	59	143000	1
##	222	35	91000	1
##	224	60	102000	1
##	225	35	60000	0
##	227	36	126000	1
##	228	56	133000	1
##	229	40	72000	0
##	230	42	80000	1
##	231	35	147000	1
##	232	39	42000	0
##	233	40	107000	1
##	235	38	112000	0
##	236	46	79000	1
##	237	40	57000	0
##	239	46	82000	0
##	240	53	143000	1
##	241	42	149000	1
##	242	38	59000	0
##	243	50	88000	1
##	244	56	104000	1
##	245	41	72000	0
##	248	57	122000	1
##	249	41	52000	0
##	252	37	52000	0
##	253	48	134000	1
##	254	37	146000	1
##	255	50	44000	0
##	256	52	90000	1
##	257	41	72000	0
##	258	40	57000	0
##	259	58	95000	1
##	261	35	77000	0
##	263	55	125000	1
##	264	35	72000	0
##	265	48	90000	1
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##	267	40	75000	0
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##	269	47	144000	1
##	270	40	61000	0
##	271	43	133000	0
##	273	60	42000	1
##	274	39	106000	1
##	275	57	26000	1
##	276	57	74000	1
##	282	35	61000	0
##	284	52	21000	1
##	290	37	78000	1
##	291	39	134000	1

##	293	55	39000	1
##	294	37	77000	0
##	295	35	57000	0
##	297	42	73000	1
##	298	43	112000	1
##	300	46	117000	1
##	301	58	38000	1
##	304	37	79000	1
##	305	40	60000	0
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##	308	47	113000	1
##	309	36	125000	1
##	310	38	50000	0
##	311	42	70000	0
##	312	39	96000	1
##	314	49	141000	1
##	315	39	79000	0
##	316	39	75000	1
##	317	54	104000	1
##	318	35	55000	0
##	319	45	32000	1
##	320	36	60000	0
##	322	53	82000	1
##	323	41	52000	0
##	324	48	30000	1
##	327	41	72000	0
##	328	42	75000	0
##	329	36	118000	1
##	331	38	51000	0
##	332	48	119000	1
##	333	42	65000	0
##	334	40	65000	0
##	336	36	54000	0
##	337	58	144000	1
##	338	35	79000	0
##	339	38	55000	0
##	341	53	104000	1
##	342	35	75000	0
##	343	38	65000	0
##	344	47	51000	1
##	345	47	105000	1
##	346	41	63000	0
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##	349	39	77000	0
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##	354	37	57000	0
##	355	36	99000	1
##	356	60	34000	1
##	357	54	70000	1
##	358	41	72000	0
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##	360	42	54000	0

##	361	43	129000	1
##	362	53	34000	1
##	364	42	79000	0
##	365	42	104000	1
##	366	59	29000	1
##	367	58	47000	1
##	368	46	88000	1
##	369	38	71000	0
##	371	60	46000	1
##	372	60	83000	1
##	374	59	130000	1
##	376	46	32000	1
##	378	42	53000	0
##	379	41	87000	1
##	381	42	64000	0
##	382	48	33000	1
##	383	44	139000	1
##	386	56	60000	1
##	388	39	71000	0
##	389	47	34000	1
##	390	48	35000	1
##	392	47	23000	1
##	393	45	45000	1
##	394	60	42000	1
##	395	39	59000	0
##	396	46	41000	1
##	397	51	23000	1
##	398	50	20000	1
##	399	36	33000	0
##	400	49	36000	1

test\_set

##	Age	EstimatedSalary	Purchased	
##	4	27	57000	0
##	6	27	58000	0
##	7	27	84000	0
##	16	29	80000	0
##	18	45	26000	1
##	22	47	49000	1
##	26	47	20000	1
##	31	31	74000	0
##	34	28	44000	0
##	35	27	90000	0
##	43	35	108000	0
##	50	31	89000	0
##	52	18	44000	0
##	56	24	55000	0
##	58	28	79000	0
##	61	27	20000	0
##	65	59	83000	0
##	68	23	82000	0
##	78	22	27000	0
##	85	30	62000	0

## 88	28	85000	0
## 90	35	50000	0
## 94	29	28000	0
## 95	29	83000	0
## 97	35	25000	0
## 99	35	73000	0
## 105	19	21000	0
## 114	37	55000	0
## 115	42	80000	0
## 116	40	57000	0
## 119	40	59000	0
## 124	35	53000	0
## 129	30	17000	0
## 131	31	58000	0
## 142	18	68000	0
## 146	24	89000	0
## 157	46	59000	0
## 160	32	135000	1
## 163	37	33000	0
## 173	26	118000	0
## 175	34	72000	0
## 183	32	117000	1
## 188	33	41000	0
## 190	28	32000	0
## 191	24	84000	0
## 198	20	36000	0
## 199	26	80000	0
## 202	49	74000	0
## 206	47	47000	0
## 210	46	22000	0
## 215	47	43000	0
## 218	40	78000	0
## 221	41	80000	0
## 223	37	144000	1
## 226	37	53000	0
## 234	49	86000	1
## 238	37	80000	0
## 246	51	146000	1
## 247	35	50000	0
## 250	35	97000	1
## 251	44	39000	0
## 260	45	131000	1
## 262	36	144000	1
## 272	59	76000	1
## 277	38	71000	0
## 278	49	88000	1
## 279	52	38000	1
## 280	50	36000	1
## 281	59	88000	1
## 283	37	70000	1
## 285	48	141000	0
## 286	37	93000	1
## 287	37	62000	0
## 288	48	138000	1

```
## 289 41          79000      0
## 292 49          89000      1
## 296 36          63000      0
## 299 45          79000      0
## 302 48          74000      1
## 303 37         137000      1
## 313 38          50000      0
## 321 52         138000      1
## 325 48         131000      1
## 326 41          60000      0
## 330 47         107000      1
## 335 57          60000      1
## 340 39         122000      1
## 347 53          72000      1
## 350 38          61000      0
## 353 42          90000      1
## 363 47          50000      1
## 370 54          26000      1
## 373 39          73000      0
## 375 37          80000      0
## 377 46          74000      0
## 380 58          23000      1
## 384 49          28000      1
## 385 57          33000      1
## 387 49          39000      1
## 391 48          33000      1
```

#feature scaling

```
training_set[-3]= scale(training_set[-3])
test_set[-3]=scale(test_set[-3])
```

#fitting KNN to the training set and predicting the test result

```
library(class)
y_pred = knn(train = training_set[,-3],
             test = test_set[,-3],
             cl = training_set[,3],
             k = 5,
             prob = TRUE)
y_pred
```

```
## [1] 0 0 0 0 1 1 1 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
## [38] 1 0 1 0 1 0 0 0 0 0 1 1 1 1 0 0 1 0 1 0 1 1 1 1 1 0 1 1 1 1 0 1
## [75] 0 1 0 0 1 1 0 1 1 0 1 1 1 1 0 1 1 1 0 0 0 1 1 1 1 1
## attr(,"prob")
## [1] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 0.8000000 1.0000000
## [8] 1.0000000 1.0000000 1.0000000 0.6000000 1.0000000 1.0000000 1.0000000
## [15] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [22] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [29] 0.8000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000
## [36] 1.0000000 0.8000000 1.0000000 1.0000000 0.8000000 1.0000000 0.6000000
## [43] 1.0000000 1.0000000 1.0000000 1.0000000 1.0000000 0.6000000 1.0000000
```



```
## [50] 1.0000000 1.0000000 0.8333333 0.8000000 1.0000000 1.0000000 0.8000000
## [57] 0.6000000 0.8000000 1.0000000 0.6000000 0.6000000 0.8000000 1.0000000
## [64] 1.0000000 1.0000000 0.8000000 0.8000000 1.0000000 1.0000000 1.0000000
## [71] 0.8000000 0.8000000 1.0000000 0.8000000 0.8000000 0.8000000 1.0000000
## [78] 0.6000000 0.6000000 1.0000000 1.0000000 0.8000000 0.8000000 1.0000000
## [85] 0.8000000 1.0000000 1.0000000 0.8000000 1.0000000 0.6000000 0.8000000
## [92] 1.0000000 1.0000000 0.6000000 0.6000000 1.0000000 1.0000000 1.0000000
## [99] 1.0000000 1.0000000
## Levels: 0 1
```

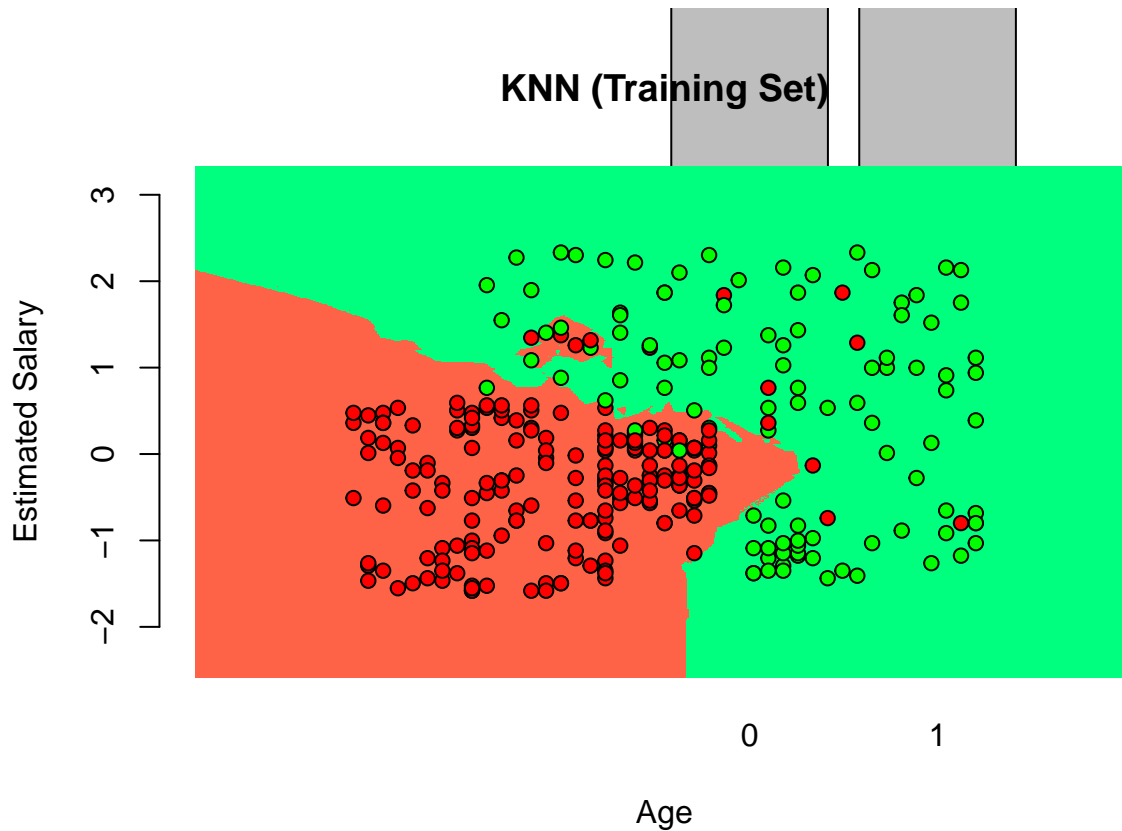
## confusion matrix

```
cm = table(test_set[,3],y_pred)
cm
```

```
##      y_pred
##      0  1
## 0 56  8
## 1  1 35
```

#visualisation - Train set

```
set = training_set
X1 = seq(min(set[,1])-1, max(set[,1])+1,0.01)
X2 = seq(min(set[,2])-1, max(set[,2])+1,0.01)
grid_set = expand.grid(X1,X2)
colnames(grid_set)= c('Age','Estimated Salary')
y_grid = knn(train= training_set[,-3],test=grid_set,cl=training_set[,3],k=5)
plot(set[,3],
      main = 'KNN (Training Set)',
      xlab = 'Age',
      ylab = 'Estimated Salary',
      xlim = range(X1),
      ylim = range(X2))
contour(X1,X2,matrix(as.numeric(y_grid),length(X1),length(X2)), add=TRUE)
points(grid_set,pch = '.', col = ifelse(y_grid == 1,'springgreen','tomato'))
points(set,pch = 21, bg = ifelse(set[,3]==1,'green','red'))
```



#Visualisation - Test dataset

```
set = test_set
X1 = seq(min(set[,1])-1, max(set[,1])+1,0.01)
X2 = seq(min(set[,2])-1, max(set[,2])+1,0.01)
grid_set = expand.grid(X1,X2)
colnames(grid_set)= c('Age','Estimated Salary')
y_grid = knn(train= training_set[,-3],test=grid_set,cl=training_set[,3],k=5)
plot(set[,3],
      main = 'KNN (Test Set)',
      xlab = 'Age',
      ylab = 'Estimated Salary',
      xlim = range(X1),
      ylim = range(X2))
contour(X1,X2,matrix(as.numeric(y_grid),length(X1),length(X2)), add=TRUE)
points(grid_set,pch = '.', col = ifelse(y_grid == 1,'springgreen','tomato'))
points(set,pch = 21, bg = ifelse(set[,3]==1,'green','red'))
```

