

eda_offender_profile_raw_data

December 11, 2021

```
[1]: import numpy as np
import pandas as pd
import altair as alt
from altair_saver import save
```

Download the offender_profile data set

```
[2]: !python download_data.py --url="http://www.csc-scc.gc.ca/005/
↳opendata-donneesouvertes/Open%20Data%20File%2020170409%20v3%20(English).csv"
↳--file_path="../data/raw/offender_profile.csv"
```

```
[3]: df_init = pd.read_csv('../data/raw/offender_profile.csv', sep=r'\s*,\s*',
header=1, encoding='ascii', engine='python')

alt.data_transformers.disable_max_rows() #normally, altair limits it to 5K rows
alt.renderers.enable('png')
```

```
[3]: RendererRegistry.enable('png')
```

```
[4]: df = df_init
df
```

```
[4]:
```

	Number	Race	Race Grouping	Gender	Age	\
0	1	WHITE	Non Indigenous	MALE	35	
1	2	WHITE	Non Indigenous	MALE	59	
2	3	METIS	Indigenous	MALE	39	
3	4	NORTH AMERICAN	Indigenous	MALE	20	
4	5	NORTH AMERICAN	Indigenous	MALE	47	
...
22765	22766	NORTH AMERICAN	Indigenous	FEMALE	58	
22766	22767	NORTH AMERICAN	Indigenous	MALE	37	
22767	22768	WHITE	Non Indigenous	MALE	50	
22768	22769	WHITE	Non Indigenous	MALE	35	
22769	22770	WHITE	Non Indigenous	FEMALE	62	

	Incarcerated/Supervised	Supervision Type	Sentence Type	\
0	Incarcerated	NaN	DETERMINATE	
1	Supervised	SR	DETERMINATE	

2	Supervised	FP	DETERMINE
3	Incarcerated	NaN	DETERMINE
4	Supervised	FP	INDETERMINE
...
22765	Incarcerated	NaN	INDETERMINE
22766	Supervised	FP	DETERMINE
22767	Incarcerated	NaN	INDETERMINE
22768	Supervised	FP	DETERMINE
22769	Supervised	FP	DETERMINE

	Aggregate Sentence Length	Province	Marital Status \
0	1095	BRITISH COLUMBIA	SINGLE
1	731	QUEBEC	SINGLE
2	912	ONTARIO	UNKNOWN
3	1233	MANITOBA	COMMON LAW
4	0	ONTARIO	SINGLE
...
22765	0	QUEBEC	DIVORCED
22766	912	MANITOBA	MARRIED
22767	0	ALBERTA	SINGLE
22768	1277	ALBERTA	DIVORCED
22769	1096	BRITISH COLUMBIA	MARRIED

	Religion
0	CATH NON-SPEC
1	PROT NON-SPEC
2	NaN
3	CHRI NON SPEC
4	NONE
...	...
22765	BAPTIST
22766	CATHOLIC - ROMA
22767	NONE
22768	NONE
22769	ANGLICAN(EPIS.

[22770 rows x 12 columns]

```
[5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 22770 entries, 0 to 22769
Data columns (total 12 columns):
#   Column              Non-Null Count  Dtype
---  -
0   Number              22770 non-null  int64
1   Race                22087 non-null  object
2   Race Grouping       22770 non-null  object
```

```

3   Gender                22770 non-null object
4   Age                  22770 non-null int64
5   Incarcerated/Supervised 22770 non-null object
6   Supervision Type      8638 non-null object
7   Sentence Type         22770 non-null object
8   Aggregate Sentence Length 22770 non-null int64
9   Province              22770 non-null object
10  Marital Status         22770 non-null object
11  Religion               20387 non-null object
dtypes: int64(3), object(9)
memory usage: 2.1+ MB

```

Here we can see that the following columns have missing values: - **Race**: we will not be filling it with any values such as median / prev / next etc - **Supervision Type**: we will likely drop this column anyway as supervision is set up after the sentencing, thus can not be a predictor - **Religion**: we will not be filling it with any values such as median / prev / next etc

In later stages, we will decide whether it makes sense to drop the rows with missing values. They may also be treated as “unreported” which can be inherently telling, since these columns are self reported. Further analysis is beyond the scope of EDA and will be address after Milestone 1.

```
[6]: df.describe()
```

```

[6]:          Number      Age  Aggregate Sentence Length
count  22770.000000  22770.000000          22770.000000
mean    11385.500000    41.954150          1416.379622
std      6573.277151    14.010167          1653.111098
min         1.000000    18.000000           0.000000
25%      5693.250000    30.000000           730.000000
50%     11385.500000    40.000000          1096.000000
75%     17077.750000    52.000000          1826.000000
max     22770.000000    94.000000         43761.000000

```

We don't really have meaningful number columns beyond the target (Aggregate Sentence Length). Almost all our features are categorical or binary.

```
[7]: df.head()
```

```

[7]:   Number      Race  Race Grouping Gender  Age  Incarcerated/Supervised \
0      1      WHITE  Non Indigenous  MALE   35      Incarcerated
1      2      WHITE  Non Indigenous  MALE   59      Supervised
2      3      METIS   Indigenous  MALE   39      Supervised
3      4  NORTH AMERICAN   Indigenous  MALE   20      Incarcerated
4      5  NORTH AMERICAN   Indigenous  MALE   47      Supervised

   Supervision Type  Sentence Type  Aggregate Sentence Length \
0              NaN    DETERMINATE          1095
1              SR    DETERMINATE           731
2              FP    DETERMINATE           912

```

3	NaN	DETERMINATE	1233
4	FP	INDETERMINATE	0

	Province	Marital Status	Religion
0	BRITISH COLUMBIA	SINGLE	CATH NON-SPEC
1	QUEBEC	SINGLE	PROT NON-SPEC
2	ONTARIO	UNKNOWN	NaN
3	MANITOBA	COMMON LAW	CHRI NON SPEC
4	ONTARIO	SINGLE	NONE

```
[8]: df.tail()
```

```
[8]:
```

	Number	Race	Race Grouping	Gender	Age \
22765	22766	NORTH AMERICAN	Indigenous	FEMALE	58
22766	22767	NORTH AMERICAN	Indigenous	MALE	37
22767	22768	WHITE	Non Indigenous	MALE	50
22768	22769	WHITE	Non Indigenous	MALE	35
22769	22770	WHITE	Non Indigenous	FEMALE	62

	Incarcerated/Supervised	Supervision Type	Sentence Type \
22765	Incarcerated	NaN	INDETERMINATE
22766	Supervised	FP	DETERMINATE
22767	Incarcerated	NaN	INDETERMINATE
22768	Supervised	FP	DETERMINATE
22769	Supervised	FP	DETERMINATE

	Aggregate Sentence Length	Province	Marital Status \
22765	0	QUEBEC	DIVORCED
22766	912	MANITOBA	MARRIED
22767	0	ALBERTA	SINGLE
22768	1277	ALBERTA	DIVORCED
22769	1096	BRITISH COLUMBIA	MARRIED

	Religion
22765	BAPTIST
22766	CATHOLIC - ROMA
22767	NONE
22768	NONE
22769	ANGLICAN(EPIS.

```
[9]: df = df_init.loc[df_init['Sentence Type'] == 'DETERMINATE']
```

```
[10]: sentence_by_prov = df.groupby('Province')['Aggregate Sentence Length'].median().
      ↪sort_values().index.tolist()

ratio_boxplots = alt.Chart(df).mark_boxplot().encode(
    x = alt.X('Aggregate Sentence Length', type='quantitative'),
```

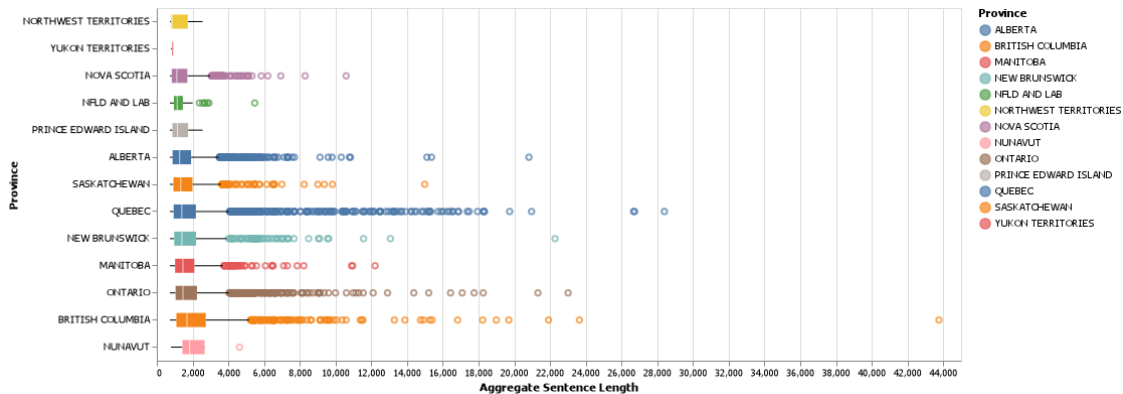
```

    y = alt.Y('Province', sort = sentence_by_prov),
    color = 'Province'
).properties(
    height = 350,
    width = 800
)

ratio_boxplots

```

[10]:



```

[11]: sentence_by_mar = df.groupby('Marital Status')['Aggregate Sentence Length'].
    ↪median().sort_values().index.tolist()

```

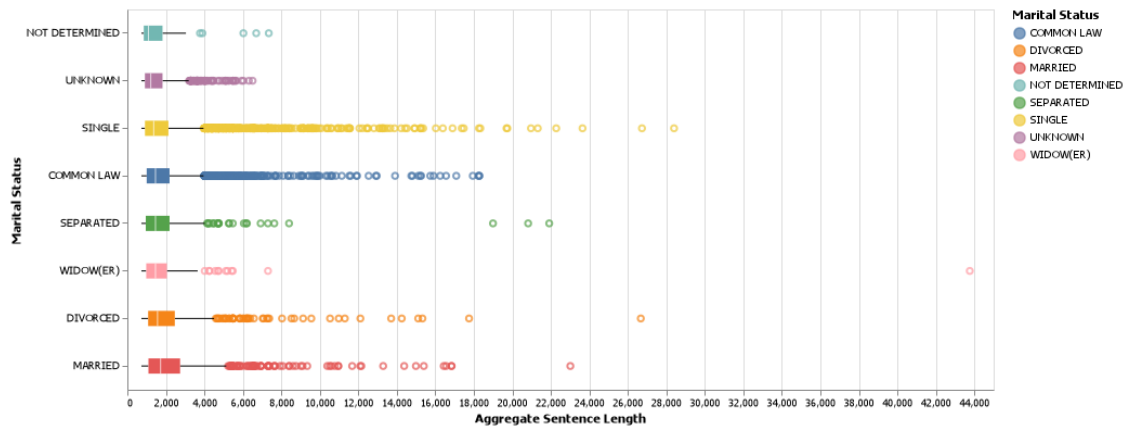
```

ratio_boxplots = alt.Chart(df).mark_boxplot().encode(
    x = alt.X('Aggregate Sentence Length', type='quantitative'),
    y = alt.Y('Marital Status', sort = sentence_by_mar),
    color = 'Marital Status'
).properties(
    height = 350,
    width = 800
)

ratio_boxplots

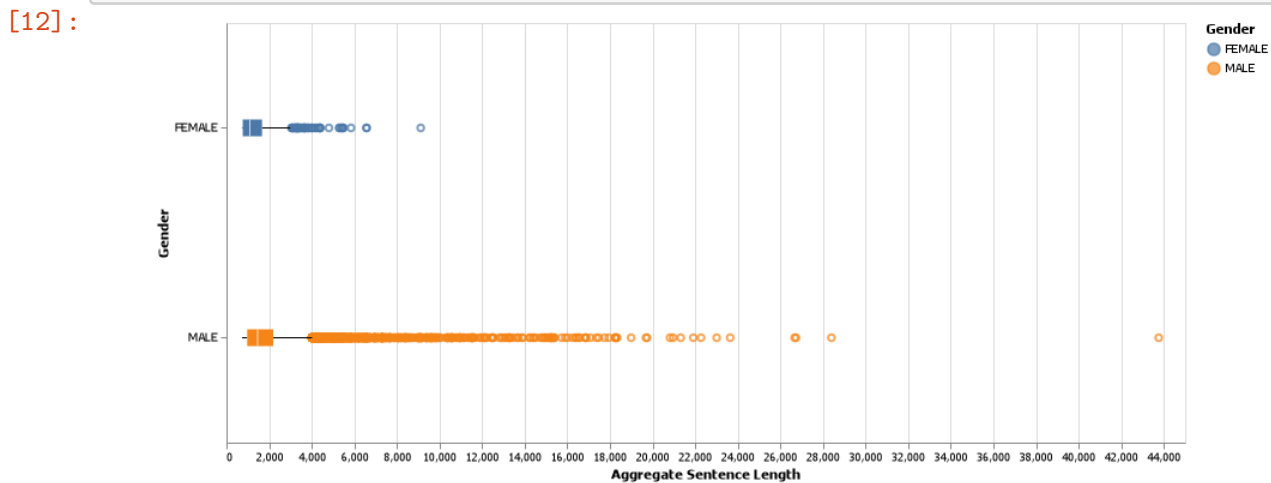
```

[11]:



```
[12]: ratio_boxplots = alt.Chart(df).mark_boxplot().encode(
    x = alt.X('Aggregate Sentence Length', type='quantitative'),
    y = alt.Y('Gender'),
    color = 'Gender'
).properties(
    height = 350,
    width = 800
)

ratio_boxplots
```



```
[13]: ratio_boxplots = alt.Chart(df).mark_boxplot().encode(
    x = alt.X('Aggregate Sentence Length', type='quantitative'),
    y = alt.Y('Race Grouping'),
    color = 'Race Grouping'
).properties(
```

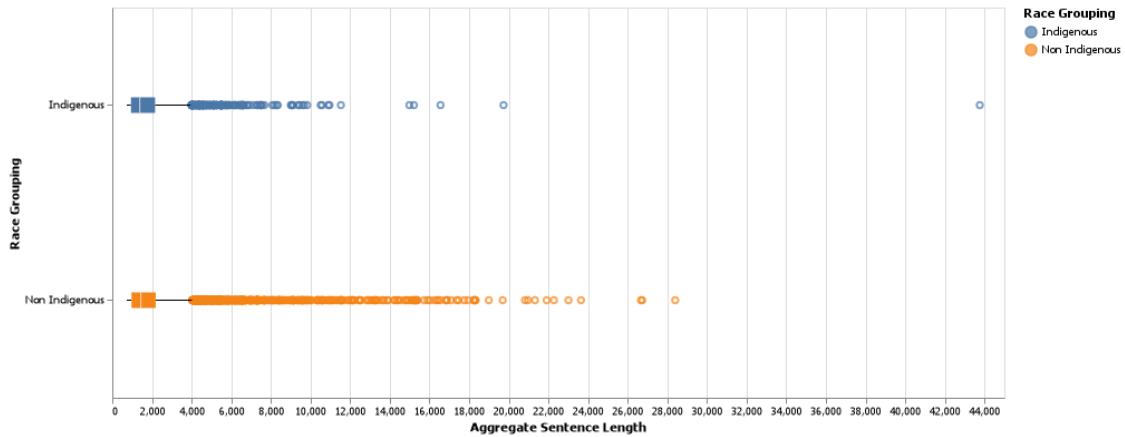
```

    height = 350,
    width = 800
)

ratio_boxplots

```

[13]:



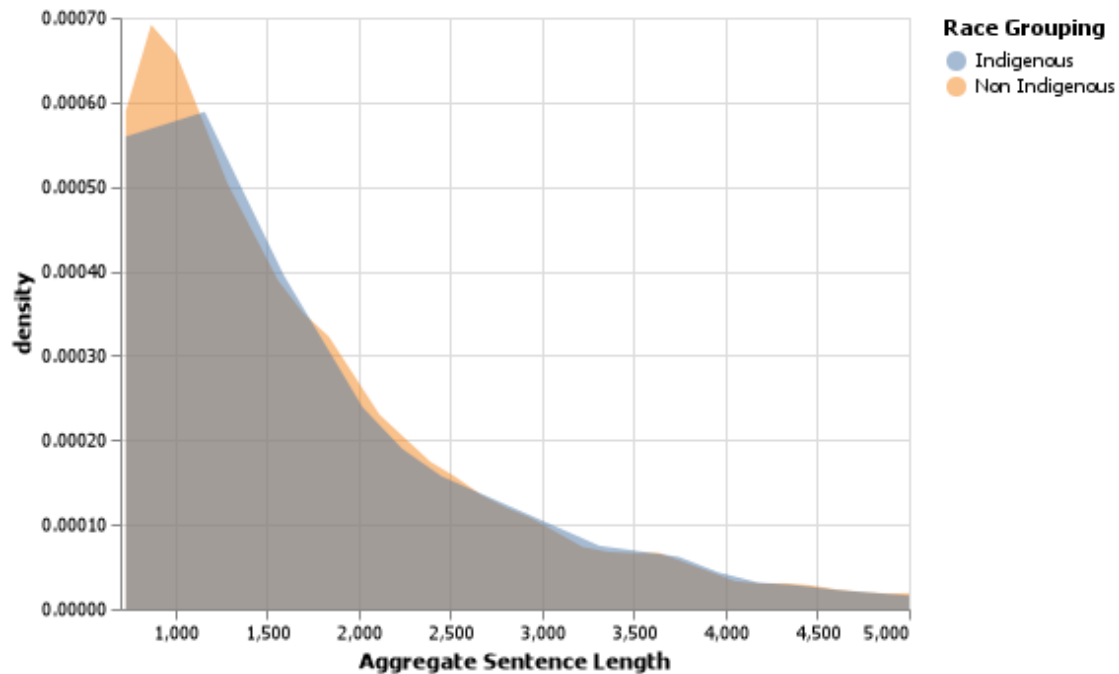
```

[14]: ratio_violins = alt.Chart(df).transform_density(
    'Aggregate Sentence Length',
    groupby=['Race Grouping'],
    as_=['Aggregate Sentence Length', 'density'],
).mark_area(
    opacity=0.5,
    clip=True
).encode(
    x=alt.X('Aggregate Sentence Length', scale=alt.Scale(domain=[700, 5000])),
    y='density:Q',
    color='Race Grouping'
)

ratio_violins

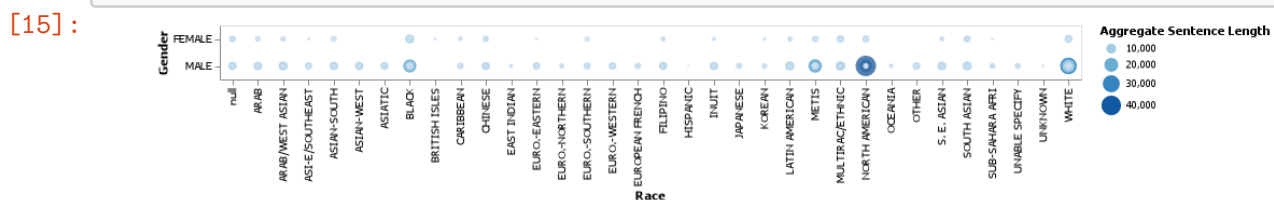
```

[14]:



```
[15]: race_gender_count_chart = alt.Chart(df).mark_circle().encode(
    x = alt.X('Race'),
    y = alt.Y('Gender'),
    size = alt.Size('Aggregate Sentence Length'),
    color = alt.Color('Aggregate Sentence Length')
).properties(
    height = 50,
    width = 800
)

# Show the plot
race_gender_count_chart
```



```
[16]: print(df['Religion'].value_counts())
```

CATH NON-SPEC 4006

NONE	2569
CATHOLIC - ROMA	1524
CHRI NON SPEC	1350
MUSLIM	1060
NATIVE SPIRITUA	936
NOT STATED	761
PROT NON-SPEC	394
BUDDHIST	340
ANGLICAN(EPIS.	316
ATHEIST	215
OTHER	195
BAPTIST	147
SIKH	127
JUDAISM	116
RASTAFARIAN	114
UNITED CHURCH	91
CHRISTIAN REFOR	79
WICCA	71
CHRISTIAN & MIS	59
SEVENTH DAY ADV	57
PENTECOSTAL ASS	57
CATHOLIC - NATI	53
PAGAN	53
AGNOSTIC	48
HINDU	47
LUTHERAN	46
CATHOLIC - GREE	43
GREEK ORTHODOX	39
JEHOVAH'S WITNE	39
PRESBYTERIAN	34
PENTECOSTAL (4	29
NAT SPIRIT PROT	24
MORMON (LATTER	22
MENNONITE	15
SALVATION ARMY	14
CHRIST. SCIENCE	11
SUFIISM	10
RUSSIAN ORTHODO	9
MAHAYANA BUDD	6
DRUIDRY PAGAN	6
ASATRUAR PAGAN	6
INDEPEND SPIRIT	6
UKRANIAN ORTHOD	5
MORAVIAN	4
METHODIST CHRIS	3
TAOISM	3
NAZARENE CHRIST	3
BAHA'I	3

PHILAD.CHUR.GOD	2
UNITARIAN	2
KRISHNA	2
HUTTERITE	1
WORLDWIDE CHURC	1
SCIENTOLOGY	1
SIDDHA YOGA	1
QUAKER SOCIETY	1
CATHOLIC - UKRA	1

Name: Religion, dtype: int64

```
[17]: print(df['Race'].value_counts())
```

WHITE	9678
NORTH AMERICAN	2731
BLACK	1346
METIS	1180
LATIN AMERICAN	201
S. E. ASIAN	168
ARAB	163
INUIT	161
ASI-E/SOUTHEAST	151
MULTIRAC/ETHNIC	140
ARAB/WEST ASIAN	125
ASIAN-SOUTH	100
SOUTH ASIAN	87
CHINESE	86
FILIPINO	67
CARIBBEAN	65
EURO.-SOUTHERN	31
EURO.-EASTERN	30
OTHER	27
SUB-SAHARA AFRI	26
ASIAN-WEST	18
UNABLE SPECIFY	16
KOREAN	12
EURO.-WESTERN	10
ASIATIC	6
OCEANIA	5
JAPANESE	5
EURO.-NORTHERN	3
EAST INDIAN	2
EUROPEAN FRENCH	2
BRITISH ISLES	1
HISPANIC	1
UNKNOWN	1

Name: Race, dtype: int64

```
[18]: print(df['Province'].value_counts())
```

```
ONTARIO          4337
QUEBEC            4144
ALBERTA           2968
BRITISH COLUMBIA  2039
SASKATCHEWAN      1057
MANITOBA          1003
NEW BRUNSWICK      863
NOVA SCOTIA        755
NFLD AND LAB       86
PRINCE EDWARD ISLAND 23
NORTHWEST TERRITORIES 7
NUNAVUT           4
YUKON TERRITORIES 3
Name: Province, dtype: int64
```

```
[19]: print(df['Race Grouping'].value_counts())
```

```
Non Indigenous    13217
Indigenous         4072
Name: Race Grouping, dtype: int64
```

```
[20]: print(df['Marital Status'].value_counts())
```

```
SINGLE            7759
COMMON LAW        5016
UNKNOWN           1652
MARRIED           1409
DIVORCED           688
SEPARATED          545
WIDOW(ER)          123
NOT DETERMINED     97
Name: Marital Status, dtype: int64
```

```
[21]: df_age = df.groupby('Age')['Aggregate Sentence Length'].mean().reset_index()
df_age
```

```
[21]:
```

	Age	Aggregate Sentence Length
0	18	1233.777778
1	19	1078.363636
2	20	1165.615894
3	21	1276.566524
4	22	1329.578512
..
66	84	2711.600000
67	85	2216.500000
68	86	1818.000000
69	88	2190.500000

70 89 1278.000000

[71 rows x 2 columns]

```
[22]: age_sentence = alt.Chart(df_age).mark_circle().encode(  
      x = 'Age',  
      y = alt.Y('Aggregate Sentence Length'))  
age_sentence
```

[22]:

