

In [1]:  
from cmdstanpy import CmdStanModel  
import arviz as az  
import numpy as np  
import scipy.stats as stats  
import pandas as pd  
import matplotlib.pyplot as plt  
import matplotlib as mpl  
import seaborn as sns  
import plotly.express as px  
plt.style.use('ggplot') #using style ggplot  
#use /usr/local/anaconda3/envs/DataAnalytics/lib/python3.8/site-packages/seaborn/\_decorators.py:36: FutureWarning: IfProgress not found. Please update Jupyter and IPython. See https://ipywidgets.readthedocs.io/en/stable/user\_install.html  
from %matplotlib\_inline import Inline as %matplotlib\_inline

Out [2]:  
data = pd.read\_csv('1\_cars.csv')  
data.head(10)

Out [2]:

	Unnamed: 0	mark	model	generation_name	year	mileage	vol_engine	fuel	city	province	price	
0	1	0	opel	combo	gen-d-2011	2015	139568	1248	Diesel	Janki	Mazowieckie	35900
1	2	1	opel	combo	gen-d-2011	2016	31991	1499	Diesel	Katowice	Śląskie	78501
2	3	0	opel	combo	gen-d-2011	2016	278437	1698	Diesel	Brzeg	Opolskie	27000
3	4	3	opel	combo	gen-d-2011	2016	476000	1248	Diesel	Korfantów	Opolskie	30800
4	5	4	opel	combo	gen-d-2011	2014	103000	1400	CNG	Tarnowskie Góry	Śląskie	35900
5	6	5	opel	combo	gen-d-2011	2017	121203	1598	Diesel	Warszawa	Mazowieckie	51900
6	7	6	opel	combo	gen-d-2011	2017	119965	1248	Diesel	Wrocław	Dolnośląskie	44700
7	8	7	opel	combo	gen-d-2011	2016	201658	1248	Diesel	Lublin	Lubelskie	29000
8	9	8	opel	combo	gen-d-2011	2014	178666	1598	Diesel	Złotów	Wielkopolskie	28900
9	10	9	opel	combo	gen-d-2011	2015	113000	1248	Diesel	Strzyżew	Mazowieckie	34900

In [3]:  
print(data.mark.unique())

Out [3]:  
'toyota' 'skoda' 'alfa-romeo' 'chevrolet' 'citroen' 'fiat' 'honda' 'hyundai' 'kia' 'mazda' 'mini' 'mitsubishi' 'nissan' 'peugeot' 'seat' 'volvo'

In [4]:  
print(data.loc[(data['mark'] == 'bmw') | (data['mark'] == 'mercedes-benz') | (data['mark'] == 'audi')].model.unique())

Out [4]:  
['a3' '3' '1' 'a2' 'a4' 'a6-allroad' 'a5' 'a7' 'q8' 'a6' 'a6-allroad' 'a7' 'rs6' 'e-tron' 'q2' 'q3' 'q4-sportback' 'q5' 'q7' 'q8' 'x3' 'x4' 'x5' 'x6' 'x7' 'z4' 'z5' 'z6' 'z7' 'z8' 'z9' 'z10' 'z11' 'z12' 'z13' 'z14' 'z15' 'z16' 'z17' 'z18' 'z19' 'z20' 'z21' 'z22' 'z23' 'z24' 'z25' 'z26' 'z27' 'z28' 'z29' 'z30' 'z31' 'z32' 'z33' 'z34' 'z35' 'z36' 'z37' 'z38' 'z39' 'z40' 'z41' 'z42' 'z43' 'z44' 'z45' 'z46' 'z47' 'z48' 'z49' 'z50' 'z51' 'z52' 'z53' 'z54' 'z55' 'z56' 'z57' 'z58' 'z59' 'z60' 'z61' 'z62' 'z63' 'z64' 'z65' 'z66' 'z67' 'z68' 'z69' 'z70' 'z71' 'z72' 'z73' 'z74' 'z75' 'z76' 'z77' 'z78' 'z79' 'z80' 'z81' 'z82' 'z83' 'z84' 'z85' 'z86' 'z87' 'z88' 'z89' 'z90' 'z91' 'z92' 'z93' 'z94' 'z95' 'z96' 'z97' 'z98' 'z99' 'z100' 'z101' 'z102' 'z103' 'z104' 'z105' 'z106' 'z107' 'z108' 'z109' 'z110' 'z111' 'z112' 'z113' 'z114' 'z115' 'z116' 'z117' 'z118' 'z119' 'z120' 'z121' 'z122' 'z123' 'z124' 'z125' 'z126' 'z127' 'z128' 'z129' 'z130' 'z131' 'z132' 'z133' 'z134' 'z135' 'z136' 'z137' 'z138' 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'z1497' 'z1498' 'z1499' 'z1500' 'z1501' 'z1502' 'z1503' 'z1504' 'z1505' 'z1506' 'z1507' 'z1508' 'z1509' 'z1510' 'z1511' 'z1512' 'z1513' 'z1514' 'z1515' 'z1516' 'z1517' 'z1518' 'z1519' 'z1520' 'z1521' 'z1522' 'z1523' 'z1524' 'z1525' 'z1526' 'z1527' 'z1528' 'z1529' 'z1530' 'z1531' 'z1532' 'z1533' 'z1534' 'z1535' 'z1536' 'z1537' 'z1538' 'z1539' 'z1540' 'z1541' 'z1542' 'z1543' 'z1544' 'z1545' 'z1546' 'z1547' 'z1548' 'z1549' 'z1550' 'z1551' 'z1552' 'z1553' 'z1554' 'z1555' 'z1556' 'z1557' 'z1558' 'z1559' 'z1560' 'z1561' 'z1562' 'z1563' 'z1564' 'z1565' 'z1566' 'z1567' 'z1568' 'z1569' 'z1570' 'z1571' 'z1572' 'z1573' 'z1574' 'z1575' 'z1576' 'z1577' 'z1578' 'z1579' 'z1580' 'z1581' 'z1582' 'z1583' 'z1584' 'z1585' 'z1586' 'z1587' 'z1588' 'z1589' 'z1590' 'z1591' 'z1592' 'z1593' 'z1594' 'z1595' 'z1596' 'z1597' 'z1598' 'z1599' 'z1600' 'z1601' 'z1602' 'z1603' 'z1604' 'z1605' 'z1606' 'z1607' 'z1608' 'z1609' 'z1610' 'z1611' 'z1612' 'z1613' 'z1614' 'z1615' 'z1616' 'z1617' 'z1618' 'z1619' 'z1620' 'z1621' 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```
INFO:cmdstanny:compiling stan file /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/one_param_ppc_age.stan to exe file /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/one_param_ppc_age.exe
INFO:cmdstanny:compiled model executable: /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/one_param_ppc_age.exe
WARNING:cmdstanny:Stan compiler has produced 2 warnings:
--- Translating Stan model to C++ code ---
bin/stanc --o=/Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/one_param_ppc_age.hpp /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/one_param_ppc_age.stan
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/one_param_ppc_age.stan, line 3, column 7: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/one_param_ppc_age.stan, line 10, column 12: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
--- Compiling, linking C++ code ---
clang++ -std=c++11 -Wno-unknown-warning-option -Wno-tautological-compare -Wno-sign-compare -D_REENTRANT -Wno-ig-mored-attribut...
INFO:cmdstanny:cmdstan start processing
chain 1 | ██████████ | 00:00 Sampling completed
INFO:cmdstanny:cmdstan done processing.
```

```
In [57]: fig, axes = plt.subplots(1, 1, figsize=(20, 15))
axes.lines[0].data_s, price_max[1], xmin=data_s_car_age.min(), xmax=data_s_car_age.max(), lines
for i in range(100):
    axes.scatter(d_short.s_price, color='black', alpha=0.5, s=10)
    axes.set_xlabel('Car age')
    axes.set_ylabel('Price')
    fig.tight_layout()
    plt.show()
```

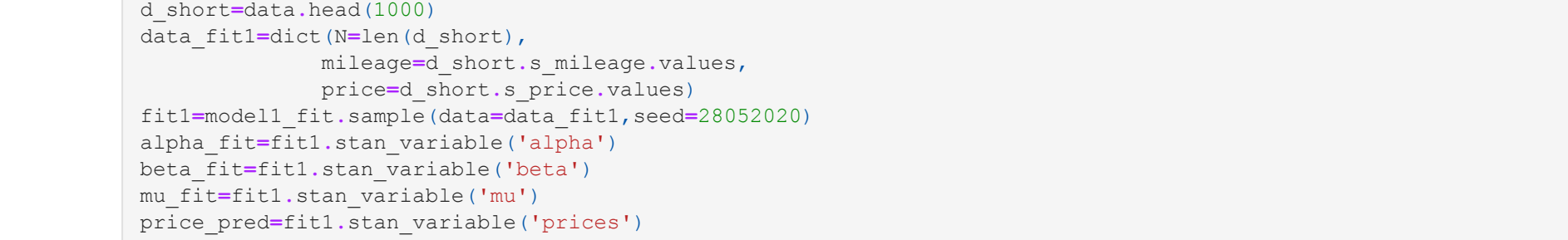


Prior data represents the data and is kept in the price min/max range good (even perfect). You can see how the prices are only decreasing in relation to the distance traveled. It's good information, so we know that prior predictive is successful.

```
In [58]: az.summary(fit2.var_names=['alpha', 'beta_car_age', 'sigma'], round_to=2, kind='stats')
```

	mean	sd	hdi	hdi_95%
<b>alpha</b>	0.70	0.50	-0.17	-0.01
<b>beta_car_age</b>	-0.38	0.01	-1.05	-1.69
<b>sigma</b>	3.20	3.24	0.00	9.39

```
In [59]: price_sims=in3.stan_variable('price')
fig, axes = plt.subplots(1, 1, figsize=(20, 10))
axes.scatter(d_short.s_price, color='black', alpha=0.5, s=10)
axes.set_xlabel('Car age')
axes.set_ylabel('Price')
fig.tight_layout()
plt.show()
```



We can verify how our predicted prices look including uncertainty model. We can see that our prior certainly assumes in very good way standard prices.

## FIRST MODEL - ONLY WITH MILEAGE PARAMETER (1' DEGREE)

```
In [81]: with open('price_fit_1.stan', 'r') as file:
data {
    int N;
    vector[N] mileage;
    real price[N];
}
parameters {
    real alpha;
    real beta;
    real lower=> sigma;
}
transformed parameters {
    vector[N] mu = alpha + beta * mileage;
}
model {
    alpha ~ normal(0.2, 0.5);
    beta ~ normal(-0.5, 0.5);
    sigma ~ exponential(0.3);
    price ~ normal(mu, sigma);
}
generated quantities {
    real prices[N];
    for (i in 1:N) {
        prices[i] = normal_rng(mu[i], sigma);
        log_lik[i] = normal_lpdf(price[i] | mu[i], sigma);
    }
}
```

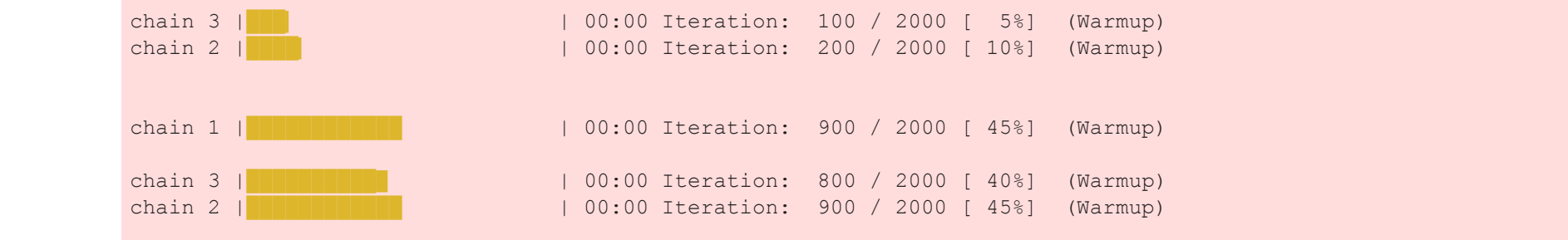
```
In [82]: model2_fit=CmdStanModel(stan_file='price_fit_1.stan')
model2_fit.compile(force=False)
d_short=data.head(1000)
data_fit=ndict({'d_short': d_short, 'mileage': d_short.s_mileage.values, 'price': d_short.s_price.values})
fit=model2_fit.sample(data_fit, seed=28052020)
alpha_fit=fit.stan_variable('alpha')
beta_fit=fit.stan_variable('beta')
mu_fit=fit.stan_variable('mu')
price_pred=fit.stan_variable('prices')
log_lik=fit.stan_variable('log_lik')
az.summary(fit2.var_names=['alpha', 'beta', 'sigma'], round_to=2, kind='stats')
```

```
INFO:cmdstanny:compiling stan file /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_1.stan to exe file /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_1.exe
INFO:cmdstanny:compiled model executable: /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_1.exe
WARNING:cmdstanny:Stan compiler has produced 3 warnings:
--- Translating Stan model to C++ code ---
bin/stanc --o=/Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_1.hpp /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_1.stan
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_1.stan, line 4, column 4: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_1.stan, line 25, column 4: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_1.stan, line 26, column 4: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
```

```
--- Compiling, linking C++ code ---
clang++ -std=c++11 -Wno-unknown-warning-option -Wno-tautological-compare -Wno-sign-compare -D_REENTRANT -Wno-ig-mored-attribut...
INFO:cmdstanny:found newer exe file, not recompiling
INFO:cmdstanny:cmdstan start processing
chain 1 | ██████████ | 00:00 Status
chain 3 | ██████████ | 00:00 Status
chain 3 | ██████████ | 00:00 Iteration: 200 / 2000 [ 10% ] (Warmup)
chain 3 | ██████████ | 00:00 Iteration: 100 / 2000 [ 5% ] (Warmup)
chain 2 | ██████████ | 00:00 Iteration: 100 / 2000 [ 5% ] (Warmup)
chain 1 | ██████████ | 00:00 Iteration: 900 / 2000 [ 45% ] (Warmup)
chain 3 | ██████████ | 00:00 Iteration: 800 / 2000 [ 40% ] (Warmup)
chain 2 | ██████████ | 00:00 Iteration: 800 / 2000 [ 40% ] (Warmup)
chain 4 | ██████████ | 00:00 Iteration: 900 / 2000 [ 45% ] (Warmup)
chain 1 | ██████████ | 00:01 Iteration: 1400 / 2000 [ 70% ] (Sampling)
chain 2 | ██████████ | 00:01 Iteration: 1400 / 2000 [ 70% ] (Sampling)
chain 4 | ██████████ | 00:01 Iteration: 1400 / 2000 [ 70% ] (Sampling)
chain 1 | ██████████ | 00:01 Iteration: 1800 / 2000 [ 90% ] (Sampling)
chain 3 | ██████████ | 00:01 Iteration: 1800 / 2000 [ 90% ] (Sampling)
chain 2 | ██████████ | 00:01 Iteration: 1800 / 2000 [ 90% ] (Sampling)
chain 1 | ██████████ | 00:02 Sampling completed
chain 2 | ██████████ | 00:02 Sampling completed
chain 4 | ██████████ | 00:02 Sampling completed
INFO:cmdstanny:cmdstan done processing.
```

	mean	sd	hdi	hdi_95%
<b>alpha</b>	-0.16	0.02	-0.19	-0.13
<b>beta</b>	-0.45	0.02	-0.49	-0.42
<b>sigma</b>	0.52	0.01	0.49	0.54

```
In [83]: fig, axes = plt.subplots(2, 1, figsize=(7, 8), sharey=True, sharex=True)
ax0=axes[0]
ax0.plot(d_short.s_mileage, d_short.s_price, color='black', alpha=0.1, s=10)
ax0.set_title('First 20 mean samples for () of datapoints'.format(len(d_short)))
ax0.set_ylabel('Price')
ax1=axes[1]
ax1.scatter(d_short.s_mileage, price_pred, ax1, supress_warning=True)
ax1.set_xlabel('Mileage')
ax1.set_ylabel('Price')
ax1.set_title('Mean uncertainty for () of datapoints'.format(len(d_short)))
fig.tight_layout()
plt.show()
```



First plot is the samples of parameters alpha and beta in order to get the standardized mean and we can see how this mean relationship gives our model thinks that is the underlying phenomena and we can see that these samples are generally consistent.

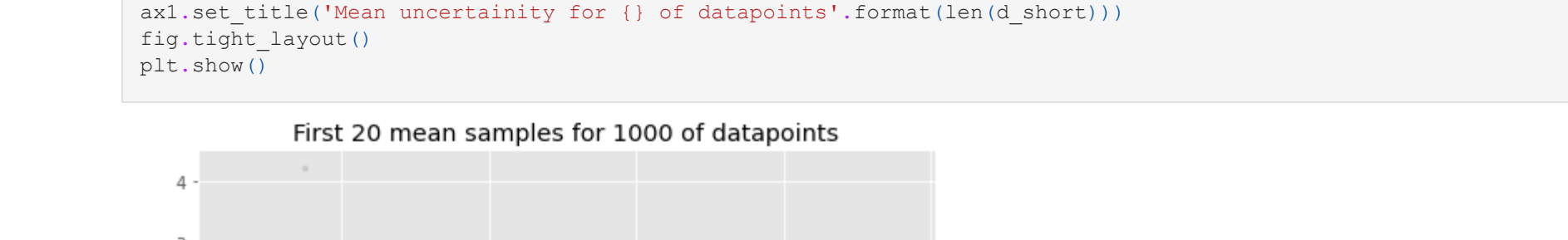
The second one shows that compute quantiles are very well concentrated.

```
In [84]: fig, axes = plt.subplots(1, 1, figsize=(7, 4))
ax=axes[0]
ax=ribbon_plot(d_short.s_mileage, price_pred, axes, supress_warning=True)
ax.set_xlabel('Mileage')
ax.set_ylabel('Price')
ax.set_title('Prediction uncertainty for () of datapoints'.format(len(d_short)))
plt.show()
```



With added the prediction uncertainty plot we can see that this prediction uncertainty is covering more and more available samples (can be better). That means all the relationship is significantly non-linear.

```
In [132]: fig, axes = plt.subplots(1, 2, figsize=(15, 10))
ax = axes[0]
ax.hist(d_short.s_price, bins=20, color='black', edgecolor='black', density=True)
ax.set_xlabel('Price')
ax.set_ylabel('Density')
ax.set_title('Histogram of Price')
ax = axes[1]
ax.hist(beta_fit, bins=20, color='black', edgecolor='black', density=True)
ax.set_xlabel('beta')
ax.set_ylabel('Density')
ax.set_title('Histogram of beta')
fig.tight_layout()
plt.show()
```



We can see that our beta and alpha are rather well concentrated. Our very white prior collapses to the very tight posterior for alpha.

## SECOND MODEL - 2 PREDICTORS (ONE WITH 2' DEGREE POLYNOMIAL FUNCTION)

Here we use higher order predictors for mileage and additional predictor for car age

```
In [66]: data['s_mileage2'] = data.s_mileage * data.s_mileage
data['s_mileage3'] = data.s_mileage * data.s_mileage * data.s_mileage
data
```

	mark	model	year	mileage	engine	fuel	price	model_types	car_age	c_mileage	s_mileage	c_car_age	
20680	audi	a7	2011	190000	3000	Gasoline	107700	F	11	14580.49846	0.148886	-0.279729	
20651	audi	a7	2012	189400	2967	Diesel	67000	F	10	13980.49846	0.142760	-0.279729	
20652	audi	a7	2010	250000	2987	Diesel	60000	F	12	14580.49846	0.142760	-0.279729	
20653	audi	a7	2016	84000	2995	Gasoline	116000	F	6	-9149.50154	0.932561	-0.279729	
20654	audi	a7	2017	90000	2995	Gasoline	149999	F	5	-85419.50154	-0.872248	-0.279729	
...	...	...	...	...	...	...	...	...	...	...	...	...	
62018	mercedes-benz	klasa	s	2018	16100	3982	Gasoline	268900	F	4	-159319.50154	-1.628666	-0.279729
62019	mercedes-benz	klasa	s	2018	37111	5482	Gasoline	325000	F	4	-138308.50154	-1.421315	-0.279729
62020	mercedes-benz	klasa	s	2017	99000	3981	Gasoline	499000	F	5	-76819.50154	-0.780346	-0.279729
62021	mercedes-benz	klasa	s	2015	14600	4663	Gasoline	258298	F	7	-14680.50154	-1.642183	-0.279729
62022	mercedes-benz	klasa	s	2017	59950	2987	Diesel	295000	F	5	-115469.50154	-1.170908	-0.279729

1623 rows x 16 columns

```
In [67]: X=data[['s_mileage', 's_mileage2']]
X
```

	s_mileage	s_mileage2
20680	0.148886	0.022167
20651	0.142760	0.020380
20652	0.142760	0.020380
20653	0.932561	0.869746
20654	-0.872248	0.760816
...	...	...
62018	-1.628666	2.648692
62019	-1.421315	1.994634
62020	-0.780346	0.606939
62021	-1.642183	2.696764
62022	-1.170908	1.390273

1623 rows x 2 columns

```
In [115]: with open('price_fit_2.stan', 'r') as file:
print(file.read())
data {
    int N;
    // number of predictions
    matrix[N, K] X; // design matrix
    vector[N] car_age;
    real price[N];
}
parameters {
    real alpha;
    vector[K] beta_mileage;
    real sigma;
}
transformed parameters {
    vector[N] mu = X * beta_mileage + car_age * (-beta_car_age) + alpha;
}
model {
    alpha ~ normal(0.2, 0.5);
    beta_mileage ~ lognormal(-1.5, 1);
    beta_car_age ~ normal(-0.2, 0.5);
    sigma ~ exponential(0.2);
    price ~ normal(mu, sigma);
}
generated quantities {
    real prices[N];
    for (i in 1:N) {
        prices[i] = normal_rng(mu[i], sigma);
        log_lik[i] = normal_lpdf(price[i] | mu[i], sigma);
    }
}
```

```
In [116]: model2_fit=CmdStanModel(stan_file='price_fit_2.stan')
model2_fit.compile(force=False)
d_short=data.head(1000)
data_fit=ndict({'d_short': d_short, 'car_age': d_short.s_car_age.values, 'mileage': d_short.s_mileage.values, 'mileage2': d_short.s_mileage2.values})
fit2=model2_fit.sample(data_fit, seed=28052020)
alpha_fit=fit2.stan_variable('alpha')
beta_mileage_fit=fit2.stan_variable('beta_mileage')
beta_car_age_fit=fit2.stan_variable('beta_car_age')
mu_fit_2=fit2.stan_variable('mu')
price_pred_2=fit2.stan_variable('prices')
log_lik_2=fit2.stan_variable('log_lik')
az.summary(fit2.var_names=['alpha', 'beta_mileage', 'beta_car_age', 'sigma'], round_to=2, kind='stats')
```

```
INFO:cmdstanny:compiling stan file /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.stan to exe file /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.exe
INFO:cmdstanny:compiled model executable: /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.exe
WARNING:cmdstanny:Stan compiler has produced 3 warnings:
--- Translating Stan model to C++ code ---
bin/stanc --o=/Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.hpp /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.stan
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.stan, line 6, column 4: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.stan, line 29, column 4: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.stan, line 30, column 4: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
Warning in /Users/kamilbaradzki/PythonProjects/ModelingofCarPrices/price_fit_2.stan, line 31, column 4: Declaration of arrays by placing brackets after a variable name is deprecated and will be removed in Stan 2.32.0. Instead use the array keyword before the type. This can be changed automatically using the auto-format flag to stanfmt.
```

```
--- Compiling, linking C++ code ---
clang++ -std=c++11 -Wno-unknown-warning-option -Wno-tautological-compare -Wno-sign-compare -D_REENTRANT -Wno-ig-mored-attribut...
INFO:cmdstanny:found newer exe file, not recompiling
INFO:cmdstanny:cmdstan start processing
chain 1 | ██████████ | 00:00 Status
chain 2 | ██████████ | 00:00 Status
chain 3 | ██████████ | 00:00 Status
chain 4 | ██████████ | 00:00 Iteration: 1 / 2000 [ 0% ] (Warmup)
chain 3 | ██████████ | 00:00 Iteration: 100 / 2000 [ 5% ] (Warmup)
chain 1 | ██████████ | 00:00 Iteration: 300 / 2000 [ 15% ] (Warmup)
chain 2 | ██████████ | 00:00 Iteration: 300 / 2000 [ 15% ] (Warmup)
chain 4 | ██████████ | 00:00 Iteration: 300 / 2000 [ 15% ] (Warmup)
chain 3 | ██████████ | 00:00 Iteration: 400 / 2000 [ 20% ] (Warmup)
chain 2 | ██████████ | 00:00 Iteration: 700 / 2000 [ 35% ] (Warmup)
chain 1 | ██████████ | 00:00 Iteration: 800 / 2000 [ 40% ] (Warmup)
chain 4 | ██████████ | 00:00 Iteration: 800 / 2000 [ 40% ] (Warmup)
chain 3 | ██████████ | 00:00 Iteration: 900 / 2000 [ 45% ] (Warmup)
chain 1 | ██████████ | 00:00 Iteration: 1001 / 2000 [ 50% ] (Sampling)
chain 3 | ██████████ | 00:00 Iteration: 1001 / 2000 [ 50% ] (Sampling)
chain 2 | ██████████ | 00:00 Iteration: 1001 / 2000 [ 50% ] (Sampling)
chain 4 | ██████████ | 00:00 Iteration: 1200 / 2000 [ 60% ] (Sampling)
chain 1 | ██████████ | 00:00 Iteration: 1400 / 2000 [ 70% ] (Sampling)
chain 3 | ██████████ | 00:00 Iteration: 1400 / 2000 [ 70% ] (Sampling)
chain 2 | ██████████ | 00:00 Iteration: 1600 / 2000 [ 80% ] (Sampling)
chain 4 | ██████████ | 00:00 Iteration: 1600 / 2000 [ 80% ] (Sampling)
chain 1 | ██████████ | 00:00 Iteration: 1700 / 2000 [ 85% ] (Sampling)
chain 3 | ██████████ | 00:00 Iteration: 1800 / 2000 [ 90% ] (Sampling)
chain 2 | ██████████ | 00:00 Iteration: 1900 / 2000 [ 95% ] (Sampling)
chain 4 | ██████████ | 00:00 Iteration: 1900 / 2000 [ 95% ] (Sampling)
chain 3 | ██████████ | 00:02 Sampling completed
chain 2 | ██████████ | 00:02 Sampling completed
chain 4 | ██████████ | 00:02 Sampling completed
INFO:cmdstanny:cmdstan done processing.
```

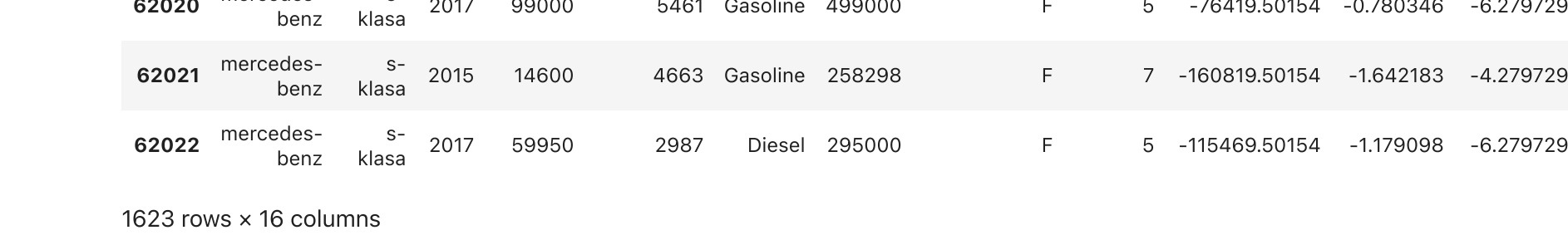
	mean	sd	hdi	hdi_95%
<b>alpha</b>	-0.28	0.02	-0.31	-0.25
<b>beta_mileage[0]</b>	-0.26	0.02	-0.29	-0.23
<b>beta_mileage[1]</b>	0.13	0.01	0.11	0.15
<b>beta_car_age</b>	0.43	0.01	0.40	0.46
<b>sigma</b>	0.37	0.01	0.36	0.39

Values are standardized so we can see that mileage variables and car age is similar important by looking at mean for these betas in stationary above

```
In [117]: log_lik
```

array[[ 2.8153e+02, -3.5354e+01, -3.8169e+02, ..., 5.50041e+03,
-1.37119e+01, 5.3437e+02,
1.71211e+04, -3.6633e+01, -3.02584e+02, ..., 9.04882e+03,
-6.70329e+02, 6.1339e+02,
1.71877e+02, -3.85074e+01, -3.17052e+02, ..., 4.67290e+07,
-1.67940e+02, 6.16272e+02,
6.38371e+02, -3.8891e+01, -4.91259e+02, ..., -4.39679e+03,
-1.49363e+01, 3.74659e+02,
1.12088e+02, -3.9719e+01, -4.89232e+02, ..., -1.42160e+02,
-1.51897e+01, 1.59033e+02,
1.11520e+03, -3.01925e+01, 2.59448e+02, ..., 5.48564e+02,
-1.69784e+05, 9.96370e+01]]

```
In [118]: fig, axes = plt.subplots(1, 1, figsize=(7, 4), sharey=True, sharex=True)
for i in range(459):
    axes.plot(d_short.s_mileage, alpha_fit[i]*beta_mileage_fit[0]+d_short.s_mileage2*beta_mileage_fit[1]+d_short.s_mileage*beta_mileage_fit[2], color='black', alpha=0.5, s=10)
    axes.set_title('Mean samples of datapoints'.format(len(d_short)))
    axes.set_ylabel('Price')
    axes.set_xlabel('Mileage')
    fig.tight_layout()
    plt.show()
```



```
In [119]: fig, axes = plt.subplots(1, 1, figsize=(7, 4), sharey=True, sharex=True)
for i in range(459):
    axes.plot(d_short.s_car_age, alpha_fit[i]*beta_car_age_fit[1]+d_short.s_car_age*beta_car_age_fit[2], color='black', alpha=0.5, s=10)
    axes.set_title('Mean samples of datapoints'.format(len(d_short)))
    axes.set_ylabel('Price')
    axes.set_xlabel('Car age')
    fig.tight_layout()
    plt.show()
```

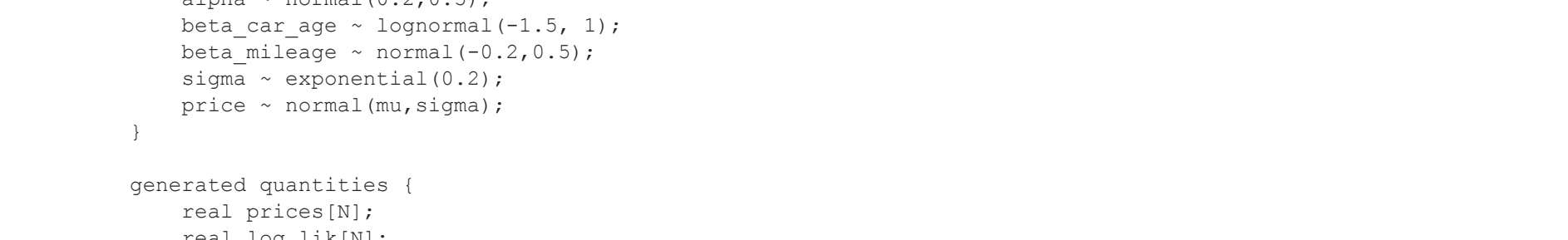


Plots above are the samples of parameters alpha and beta in order to get the standardized mean and we can see how this mean relationship gives our model thinks that is the underlying phenomena and we can see that these samples are generally consistent in that two cases.

```
In [120]: fig, axes = ribbon_plot(d_short.s_mileage, price_pred_2, axes, supress_warning=True)
ax=axes[0]
ax=scatter(d_short.s_mileage, d_short.s_price, color='black', alpha=0.2, s=10)
ax.set_xlabel('Mileage')
ax.set_ylabel('Price')
ax.set_title('Prediction uncertainty for 1000 of datapoints'.format(len(d_short)))
plt.show()
```

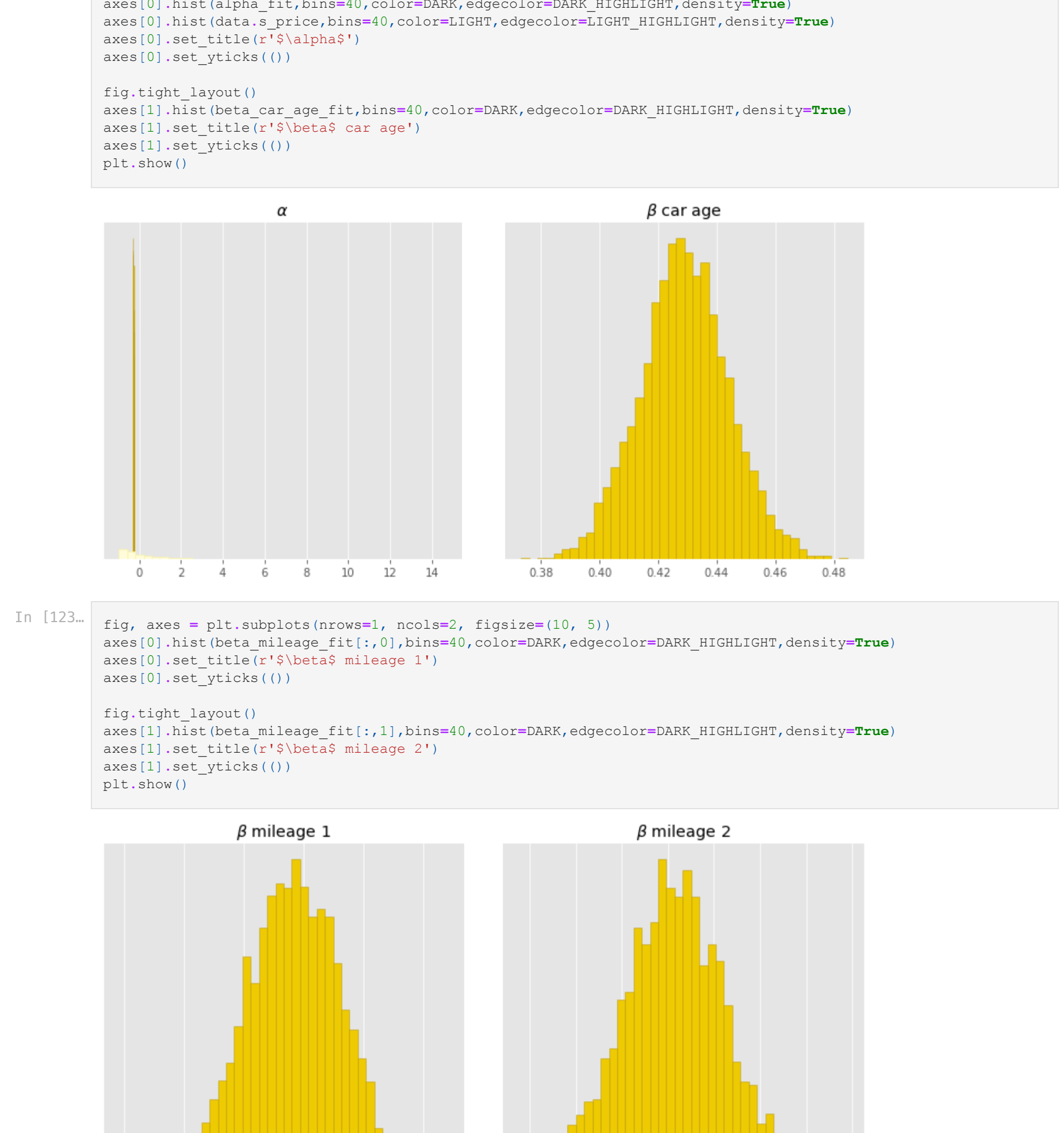


```
In [121]: fig, axes = ribbon_plot(d_short.s_mileage, price_pred_2, axes, supress_warning=True)
ax=axes[0]
ax=scatter(d_short.s_mileage, d_short.s_price, color='black', alpha=0.2, s=10)
ax.set_xlabel('Mileage')
ax.set_ylabel('Price')
ax.set_title('Prediction uncertainty for 1000 of datapoints'.format(len(d_short)))
plt.show()
```





The plots of prediction uncertainty shows that compute quantiles are very well concentrated. The higher order predictors here get even better representation of our data



### Comparing models

```
In [124]: # Comparing 2 models: model_1 - with mileage only, model_2 - with mileage and car age
idata1 = az.from_cmdstanpy(posterior = fit1, log_likelihood = "log_lik")
idata2 = az.from_cmdstanpy(posterior = fit2, log_likelihood = "log_lik")
compare_dict = {"model_1": idata1, "model_2": idata2}
az.compare(compare_dict, ic = "loo") # loo stands for leave one out cross validation
```

	rank	0	loo	p_loo	d_loo	weight	se	dse	warning	loo_scale
model_2	0	-432.320566	10.518179	0.000000	0.999509	56.588405	0.000000	False	log	
model_1	1	-767.656092	5.598634	335.335527	0.000491	40.058480	23.539642	False	log	

```
In [125]: # Comparing 2 models: model_1 - with mileage only, model_2 - with mileage and car age
idata1 = az.from_cmdstanpy(posterior = fit1, log_likelihood = "log_lik")
idata2 = az.from_cmdstanpy(posterior = fit2, log_likelihood = "log_lik")
compare_dict = {"model_1": idata1, "model_2": idata2}
az.compare(compare_dict, ic = "waic") # loo stands for leave one out cross validation

/usr/local/anaconda3/envs/DataAnalytics/lib/python3.8/site-packages/arviz/stats/state.py:1458: UserWarning: For one or more samples the posterior variance of the log predictive densities exceeds 0.4. This could be indication of WAIC starting to fail.
See http://arxiv.org/abs/1507.04544 for details
warnings.warn(
/usr/local/anaconda3/envs/DataAnalytics/lib/python3.8/site-packages/arviz/stats/state.py:1458: UserWarning: For one or more samples the posterior variance of the log predictive densities exceeds 0.4. This could be indication of WAIC starting to fail.
See http://arxiv.org/abs/1507.04544 for details
warnings.warn(
```

	rank	0	waic	p_waic	d_waic	weight	se	dse	warning	waic_scale
model_2	0	-432.339402	10.537014	0.000000	0.999509	56.606168	0.000000	False	log	
model_1	1	-767.694951	5.637493	335.35555	0.000491	40.082278	23.531161	True	log	

```
In [126]: az.plot_compare(az.compare(compare_dict))
```

