

A photograph of the front of a blue pickup truck. The hood is heavily rusted, particularly in the center. The truck is parked outdoors, with a white building on the right and trees in the background. The text "The Mission: How not to get ripped off on Craig's List" is overlaid in large black font across the upper half of the image.

The Mission: How not to get ripped off on Craig's List

Detective: Vicky Clayton



The Why:
Trying to buy a
car

cars & trucks

search cars & trucks

save search

all owner dealer

- search titles only
- has image
- posted today
- bundle duplicates
- include nearby areas

MILES FROM ZIP
miles: from zip

PRICE
min max

MAKE AND MODEL
make / model

MODEL YEAR
min max

ODOMETER
min max

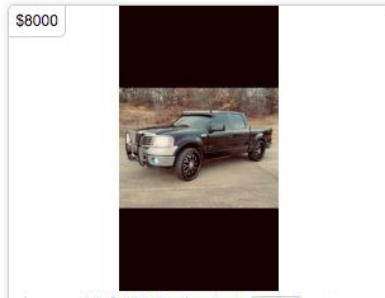
cryptocurrency ok

- condition
- cylinders
- drive
- fuel
- paint color



\$15500

★ Feb 2 Dodge Ram 1500 ST \$15500 (Eufaula ok)

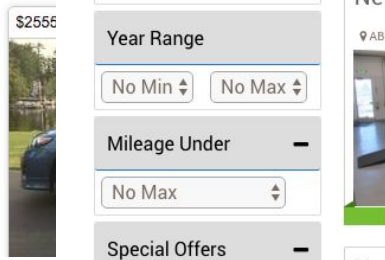


\$8000

★ Feb 2 07 F-150 Lariat 4x4 \$8000 (Fort smith)



\$850



\$2555

No Min

No Max

Year Range

No Min No Max

Mileage Under

No Max

Special Offers

- CERTIFIED PRE-OWNED (146)
- RECENTLY REDUCED (286)
- UNDER \$35K SPECIALS (102)

Make

- CHEVROLET (17)
- CHRYSLER (242)

New 2016 Honda Odyssey EX-L

ABOUT 9 MILES FROM 94103

Stock	GB093073	price: \$71,900
Mileage	322	dealer discount: \$8,000
Color	White	sale price: \$63,900
Conv Make	VMI	
Conversion	Side Entry Automatic In Floor	
Location	Oakland CA	

Recently Reduced

[MORE DETAILS](#) [REQUEST MORE INFO](#) [COMPARE](#)

New 2016 Honda Odyssey EX-L

ABOUT 9 MILES FROM 94103

Stock	GB113417	Request More Info for Price
Mileage	316	
Color	Black	
Conv Make	VMI	
Conversion	Side Entry Automatic In Floor	
Location	Oakland CA	

[MORE DETAILS](#) [REQUEST MORE INFO](#) [COMPARE](#)

N = 39000

Mileage

Age

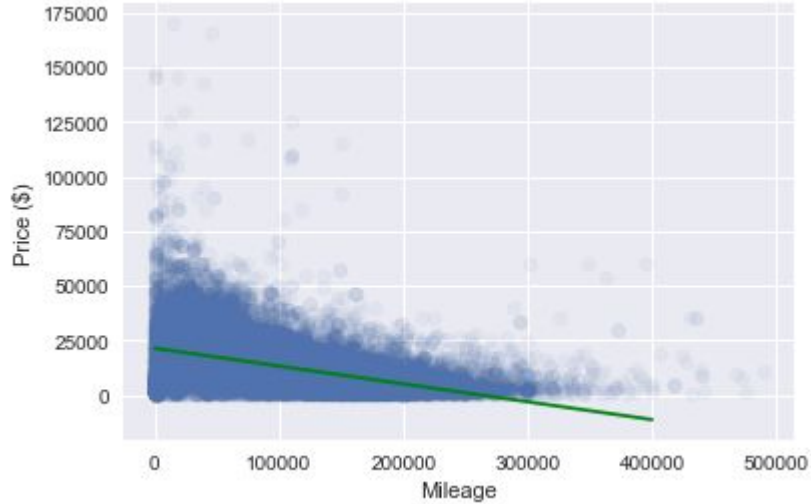
Brand

“Condition”

Paint color



Start simple...



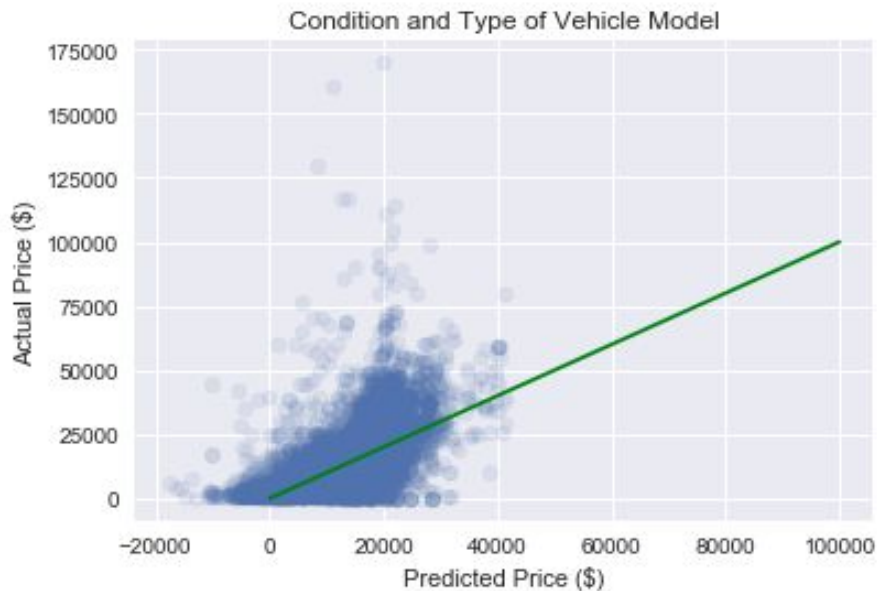
RMSE (test):

\$10948

(Mean \$: 12190)

When you don't succeed, try, try again...

RMSE (test): \$9223



RMSE (test): \$8886



Mileage	Age	Condition	Type	Paint color	Brand
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Let's look at some p-values...

const	2.345e+04	255.690	91.705	0.000	2.29e+04	2.39e+04
age	-229.1697	5.731	-39.986	0.000	-240.403	-217.936
mileage	-0.0622	0.001	-81.055	0.000	-0.064	-0.061
excellent	-1731.0408	120.371	-14.381	0.000	-1966.971	-1495.110
fair	-5632.5516	267.040	-21.093	0.000	-6155.957	-5109.146
good	-4492.2881	139.602	-32.179	0.000	-4765.911	-4218.665
like_new	1291.0620	179.208	7.204	0.000	939.810	1642.314
new	1.052e+04	558.725	18.821	0.000	9420.679	1.16e+04
SUV	1592.3041	138.560	11.492	0.000	1320.723	1863.885
bus	1.129e+04	1531.501	7.375	0.000	8292.799	1.43e+04
mini-van	26.3157	388.332	0.068	0.946	-734.824	787.456

```
('age', -0.009816733299688369),  
('mileage', -2.0708583557851332e-06),
```

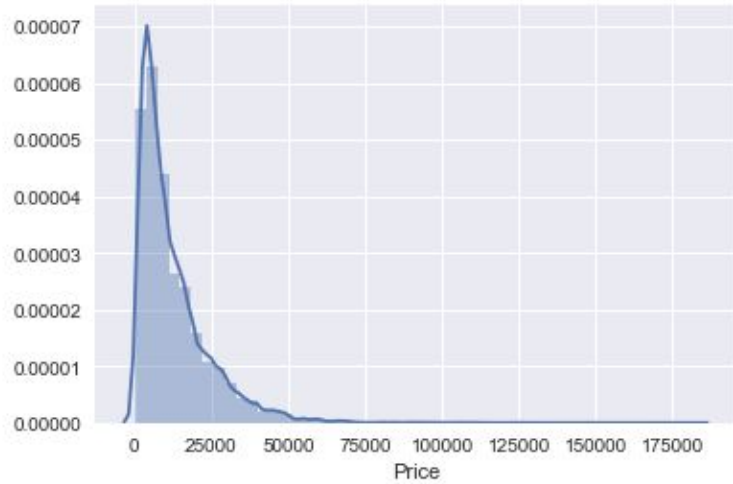
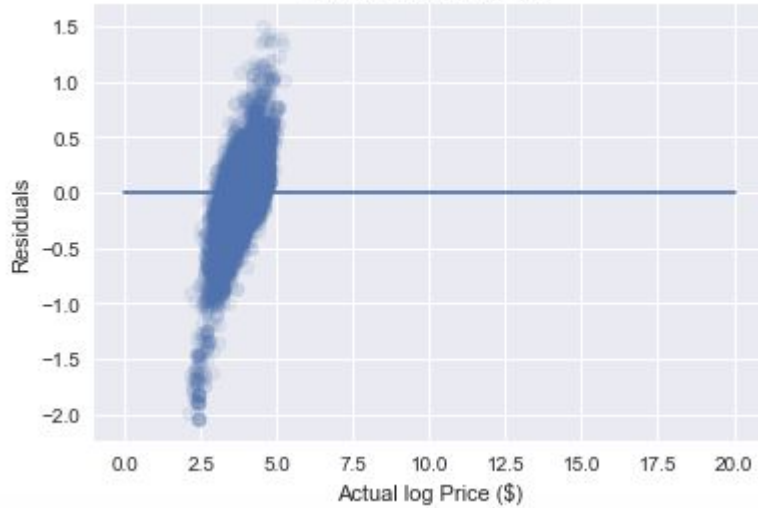
	age	mileage
age	1.000000	0.294187
mileage	0.294187	1.000000

Let's look at some p-values...

const	2.345e+04	255.690	91.705	0.000	2.29e+04	2.39e+04
age	-229.1697	5.731	-39.986	0.000	-240.403	-217.936
mileage	-0.0622	0.001	-81.055	0.000	-0.064	-0.061
excellent	-1731.0408	120.371	-14.381	0.000	-1966.971	-1495.110
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Omnibus:	29693.414	Durbin-Watson:	1.431
Prob(Omnibus):	0.000	Jarque-Bera (JB):	1395603.745
Skew:	3.231	Prob(JB):	0.00
Kurtosis:	31.583	Cond. No.	1.18e+16

Model 4: Residual Plot



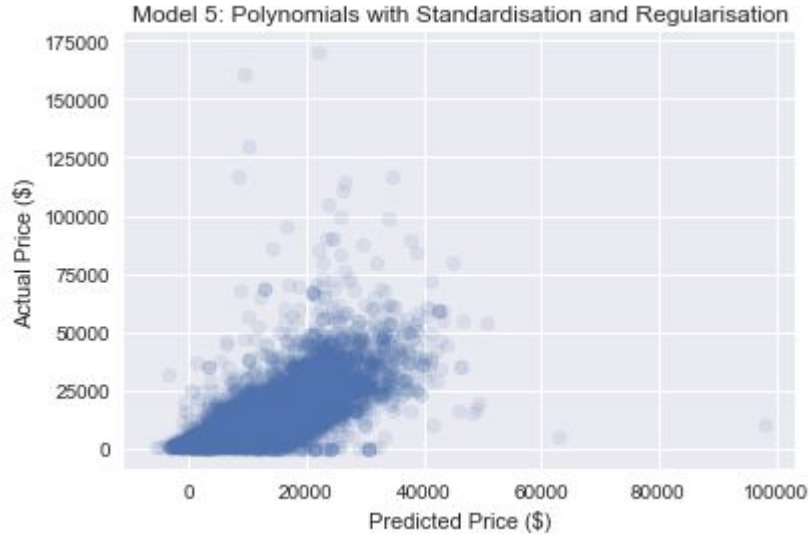
Still lots of heteroskedasticity
even after transforming it



Missing Signal.



Ok, we need to add some complexity



RMSE: \$8880

(Worse than the linear regression of order 2 with an RMSE of \$8042)



Next Steps

- Focus on one dataset with more features before doing regularisation
- Then grid search across lambdas to find optimal regularisation
- Cross-validate models to make estimates of error more robust



Or, I could just go to Kelley's Blue Book



Valid for ZIP Code 95695 through 02/08/2018

