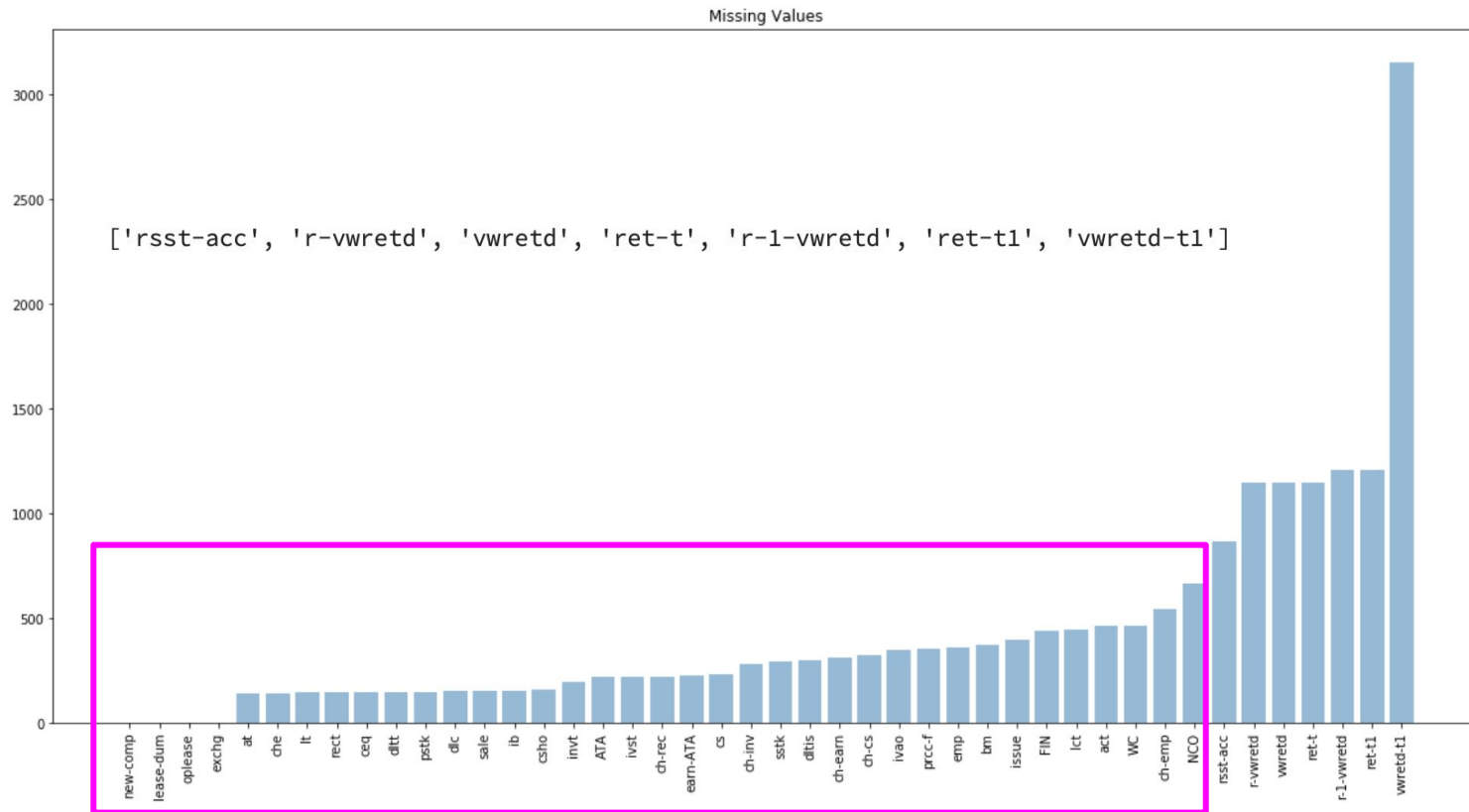


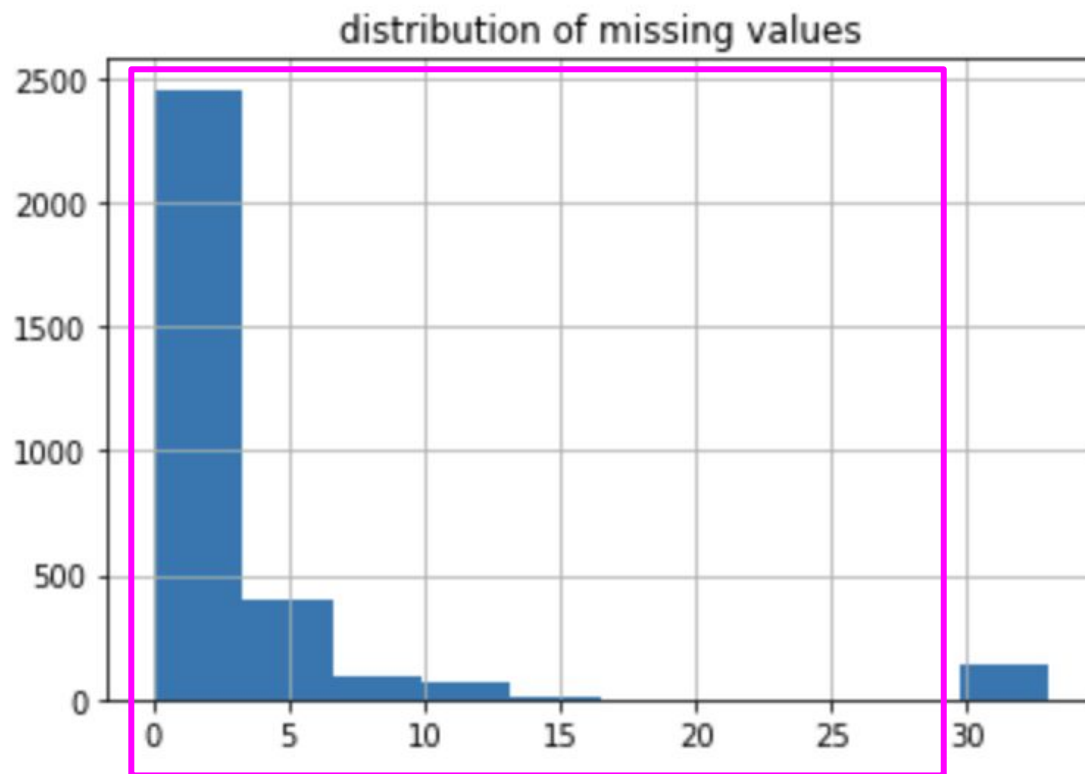
# Dataset A + SdA

Hannah LeBlanc CISC873

# Missing Values by Column



# Missing Values by Row



Data size after removing rows  
and columns

X shape: (3013, 37)

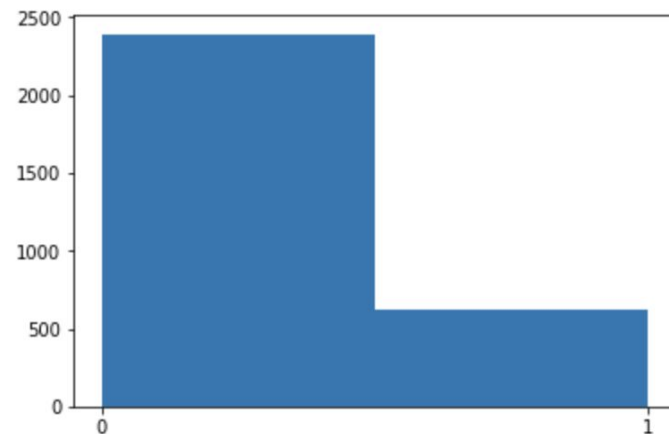
y shape: (3013, 1)

```
X_all['lease-dum'].unique()
```

```
array([0, 1])
```

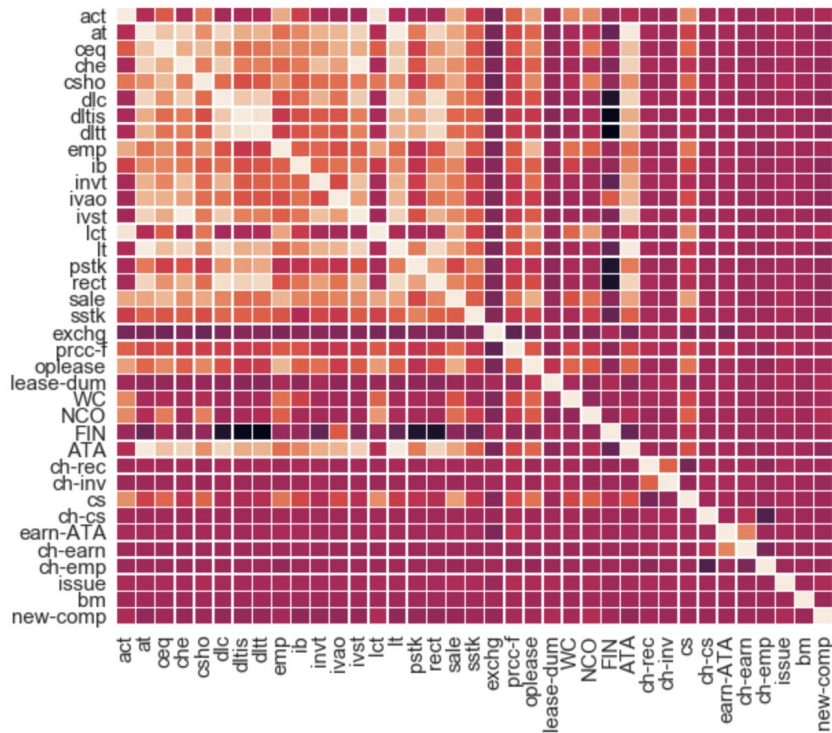
```
len(X_all['new-comp'].unique())
```

```
Data columns (total 37 columns):
act          3013 non-null float64
at           3013 non-null float64
ceq          3013 non-null float64
che          3013 non-null float64
csho         3013 non-null float64
dlc          3013 non-null float64
dltis        3013 non-null float64
dltt         3013 non-null float64
emp          3013 non-null float64
ib           3013 non-null float64
invnt        3013 non-null float64
ivao         3013 non-null float64
ivst         3013 non-null float64
lct          3013 non-null float64
lt           3013 non-null float64
pstk         3013 non-null float64
rect         3013 non-null float64
sale         3013 non-null float64
sstk         3013 non-null float64
exchg        3013 non-null int64
prcc-f       3013 non-null float64
oplease      3013 non-null float64
lease-dum    3013 non-null int64
WC           3013 non-null float64
NCO          3013 non-null float64
FIN          3013 non-null float64
ATA          3013 non-null float64
ch-rec       3013 non-null float64
ch-inv       3013 non-null float64
cs           3013 non-null float64
ch-cs        3013 non-null float64
earn-ATA     3013 non-null float64
ch-earn      3013 non-null float64
ch-emp       3013 non-null float64
issue        3013 non-null float64
bm           3013 non-null float64
new-comp     3013 non-null int64
```

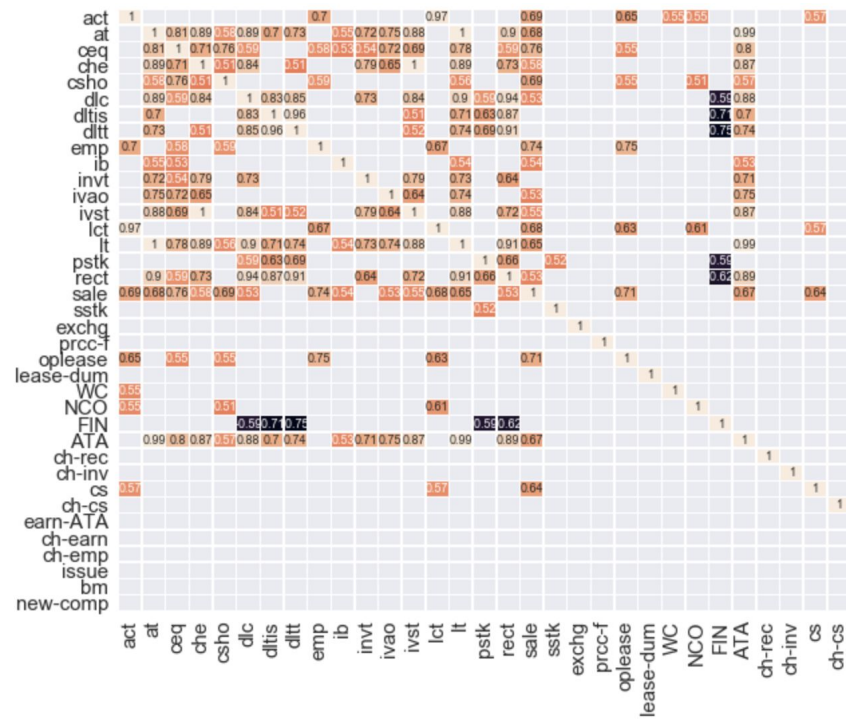


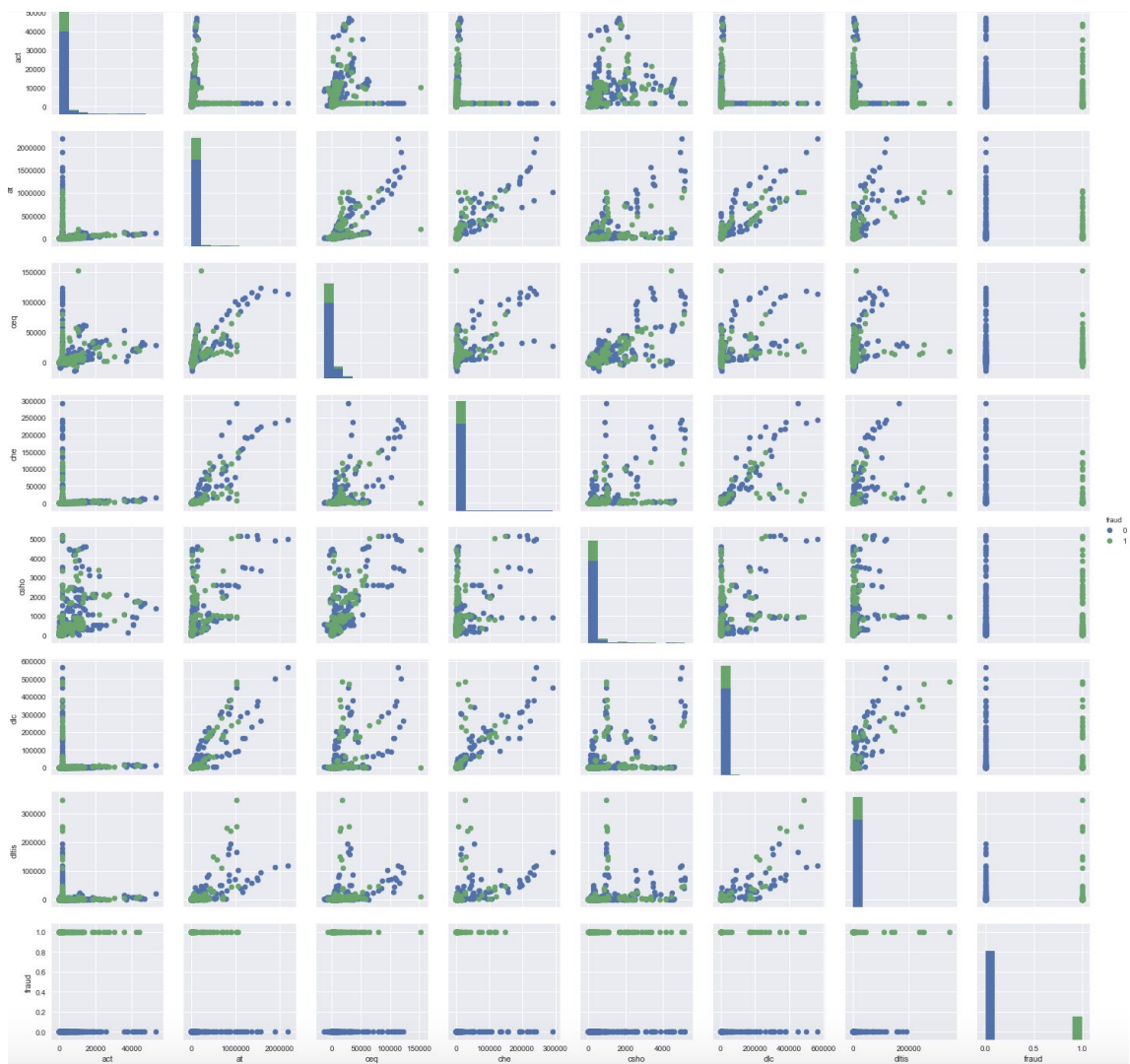
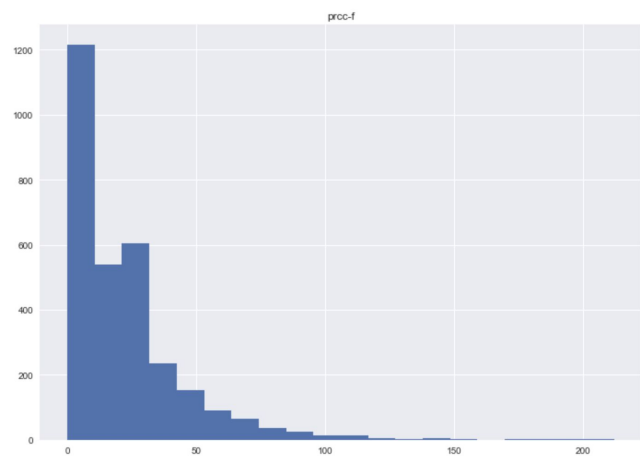
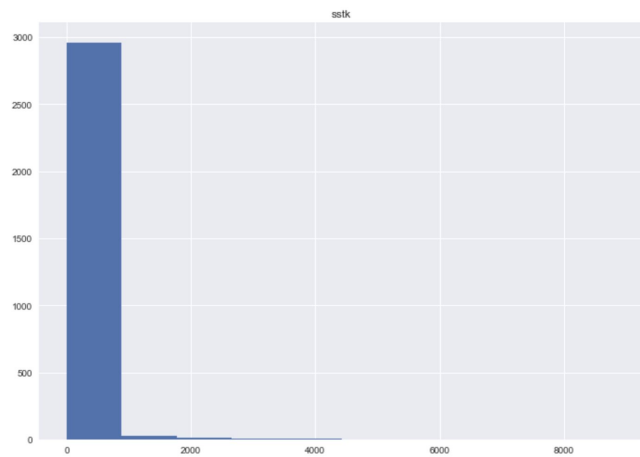
Not fraud (0), 2392

Fraud (1), 621



(corr >= 0.5) | (corr <= -0.4)





1. Train: (1807, 37)  
Validate: (603, 37)  
Test: (603, 37)

2. `transformer = MinMaxScaler(feature_range=(0, 1)).fit(X_train)`  
`X_train = transformer.transform(X_train)`  
`X_validate = transformer.transform(X_validate)`  
`X_test = transformer.transform(X_test)`

3.  
`noise_factor = 0.4`  
`x_train_noisy = X_train + noise_factor * np.random.normal(loc=0.0, scale=1.0, size=X_train.shape)`  
`x_validate_noisy = X_validate + noise_factor * np.random.normal(loc=0.0, scale=1.0, size=X_validate.shape)`  
`x_test_noisy = X_test + noise_factor * np.random.normal(loc=0.0, scale=1.0, size=X_test.shape)`

4.

```
history = ae.fit(x_train_noisy, X_train, batch_size=batch_size, epochs=epochs,  
                verbose=0, shuffle=True, validation_data=(x_validate_noisy, X_validate),  
                callbacks=[checkpointer, tensorboard]).history
```

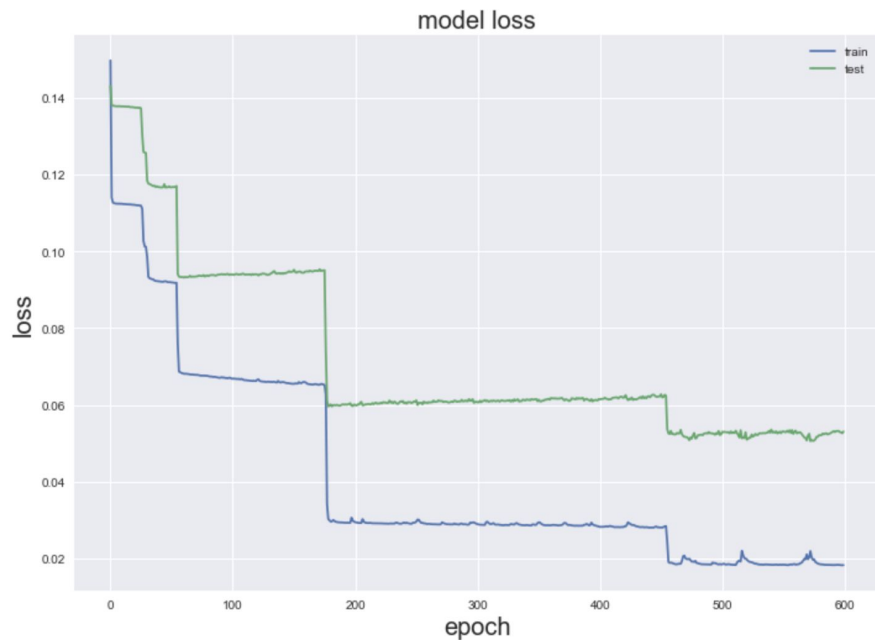
# Autoencoder

Layer (type)		Output Shape	Param #
=====		=====	=====
Encoder	input_48 (InputLayer)	(None, 37)	0
	dense_381 (Dense)	(None, 32)	1216
	dense_382 (Dense)	(None, 16)	528
	dense_383 (Dense)	(None, 8)	136
Latent Layer	dense_384 (Dense)	(None, 2)	18
Decoder	dense_385 (Dense)	(None, 8)	24
	dense_386 (Dense)	(None, 16)	144
	dense_387 (Dense)	(None, 32)	544
	dense_388 (Dense)	(None, 37)	1221
=====		=====	=====



# Train & Validation

```
ae.compile(metrics=['accuracy'], loss='mean_squared_error', optimizer='adam')
```



# Clustering with Denoising Autoencoders

Latent Layer w. Training Data

