HW1_part2

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1 CS 84020 Neural Networks and Deep Learning

Homework 1

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2 PART2

3 Listing 6

Load libraries. Load the Wisconsin Diagnostic Breast Cancer dataset. Data Visualization. Analysis The list of pairwise Pearson correlation of data shows that 'perimeter_mean' and 'radius_mean' have the largest correlation among all the 30 attributes. From the correlation matrix, we conclude that the 'perimeter', 'radius' and 'area' are highly correlated because the mean, se and worst value of these three attributes have have large correlation. On the other hand, 'texture' is less related to all of the three. 'fractal_dimension', 'texture', 'symmetry' and 'smoothness' are four features that have the least correlation with other features. The skew of data shows that 'area_se' has the most positive skew. This means 'area_se' has some extreme larger data. We can also conclude this from the density plot: For 'area_se', the tail on the right is longer than the tail on the left, because there are a few variables that have large values, which makes the tail on the right side of the curve longer.

a) Load libraries.

[1]:
```
# Load libraries
from pandas import read_csv, set_option
from pandas.plotting import scatter_matrix
from matplotlib import pyplot as plt
import seaborn as sns
import pandas as pd
import numpy as np
from google.colab import data_table
from sklearn.model_selection import train_test_split
from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
```

```
from sklearn.metrics import classification_report
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
from sklearn.naive_bayes import GaussianNB
from sklearn.svm import SVC
from sklearn.preprocessing import MinMaxScaler, StandardScaler, Normalizer,␣
 ↪Binarizer
```

[2]:
```
# Set options
set_option('display.max_columns', 32)

# plt.style.use('seaborn-talk')
plt.style.use('seaborn-white')
```

b) Load the dataset.

[3]:
```
# Load dataset
filename = 'http://archive.ics.uci.edu/ml/machine-learning-databases/
 ↪breast-cancer-wisconsin/wdbc.data'
colnames = ['id', 'diagnosis',
            'radius_mean', 'texture_mean', 'perimeter_mean', 'area_mean',
            'smoothness_mean', 'compactness_mean', 'concavity_mean',
            'concave_points_mean', 'symmetry_mean', 'fractal_dimension_mean',
            'radius_se', 'texture_se', 'perimeter_se', 'area_se',
            'smoothness_se', 'compactness_se', 'concavity_se',
            'concave_points_se', 'symmetry_se', 'fractal_dimension_se',
            'radius_worst', 'texture_worst', 'perimeter_worst', 'area_worst',
            'smoothness_worst', 'compactness_worst', 'concavity_worst',
            'concave_points_worst', 'symmetry_worst', 'fractal_dimension_worst']
dataset = read_csv(filename, names=colnames, header=None)
```

[4]:
```
# drop `id` identifier
dataset = dataset.drop(['id'], 1)

# recode target labels to 0 and 1
dataset['diagnosis'] = dataset['diagnosis'].map({'B':0, 'M':1})
dataset.head(3)
```

[4]:

	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	\
0	1	17.99	10.38	122.8	1001.0	
1	1	20.57	17.77	132.9	1326.0	
2	1	19.69	21.25	130.0	1203.0	

```
   smoothness_mean  compactness_mean  concavity_mean  concave_points_mean  \
0          0.11840           0.27760          0.3001              0.14710
1          0.08474           0.07864          0.0869              0.07017
2          0.10960           0.15990          0.1974              0.12790

   symmetry_mean  fractal_dimension_mean  radius_se  texture_se  perimeter_se  \
0         0.2419                 0.07871     1.0950      0.9053         8.589
1         0.1812                 0.05667     0.5435      0.7339         3.398
2         0.2069                 0.05999     0.7456      0.7869         4.585

   area_se  smoothness_se  compactness_se  concavity_se  concave_points_se  \
0   153.40       0.006399         0.04904       0.05373            0.01587
1    74.08       0.005225         0.01308       0.01860            0.01340
2    94.03       0.006150         0.04006       0.03832            0.02058

   symmetry_se  fractal_dimension_se  radius_worst  texture_worst  \
0      0.03003              0.006193         25.38          17.33
1      0.01389              0.003532         24.99          23.41
2      0.02250              0.004571         23.57          25.53

   perimeter_worst  area_worst  smoothness_worst  compactness_worst  \
0            184.6      2019.0            0.1622             0.6656
1            158.8      1956.0            0.1238             0.1866
2            152.5      1709.0            0.1444             0.4245

   concavity_worst  concave_points_worst  symmetry_worst  \
0           0.7119                0.2654          0.4601
1           0.2416                0.1860          0.2750
2           0.4504                0.2430          0.3613

   fractal_dimension_worst
0                  0.11890
1                  0.08902
2                  0.08758
```

c) Data visualization.

Pairwise Pearson Correlation

[5]:
```
features_pearson = list(dataset.columns[1:32])
dataset[features_pearson].corr(method='pearson')
```

[5]:
```
                         radius_mean  texture_mean  perimeter_mean  area_mean  \
radius_mean                 1.000000      0.323782        0.997855   0.987357
texture_mean                0.323782      1.000000        0.329533   0.321086
perimeter_mean              0.997855      0.329533        1.000000   0.986507
area_mean                   0.987357      0.321086        0.986507   1.000000
smoothness_mean             0.170581     -0.023389        0.207278   0.177028
```

compactness_mean	0.506124	0.236702	0.556936	0.498502
concavity_mean	0.676764	0.302418	0.716136	0.685983
concave_points_mean	0.822529	0.293464	0.850977	0.823269
symmetry_mean	0.147741	0.071401	0.183027	0.151293
fractal_dimension_mean	-0.311631	-0.076437	-0.261477	-0.283110
radius_se	0.679090	0.275869	0.691765	0.732562
texture_se	-0.097317	0.386358	-0.086761	-0.066280
perimeter_se	0.674172	0.281673	0.693135	0.726628
area_se	0.735864	0.259845	0.744983	0.800086
smoothness_se	-0.222600	0.006614	-0.202694	-0.166777
compactness_se	0.206000	0.191975	0.250744	0.212583
concavity_se	0.194204	0.143293	0.228082	0.207660
concave_points_se	0.376169	0.163851	0.407217	0.372320
symmetry_se	-0.104321	0.009127	-0.081629	-0.072497
fractal_dimension_se	-0.042641	0.054458	-0.005523	-0.019887
radius_worst	0.969539	0.352573	0.969476	0.962746
texture_worst	0.297008	0.912045	0.303038	0.287489
perimeter_worst	0.965137	0.358040	0.970387	0.959120
area_worst	0.941082	0.343546	0.941550	0.959213
smoothness_worst	0.119616	0.077503	0.150549	0.123523
compactness_worst	0.413463	0.277830	0.455774	0.390410
concavity_worst	0.526911	0.301025	0.563879	0.512606
concave_points_worst	0.744214	0.295316	0.771241	0.722017
symmetry_worst	0.163953	0.105008	0.189115	0.143570
fractal_dimension_worst	0.007066	0.119205	0.051019	0.003738

	smoothness_mean	compactness_mean	concavity_mean	\
radius_mean	0.170581	0.506124	0.676764	
texture_mean	-0.023389	0.236702	0.302418	
perimeter_mean	0.207278	0.556936	0.716136	
area_mean	0.177028	0.498502	0.685983	
smoothness_mean	1.000000	0.659123	0.521984	
compactness_mean	0.659123	1.000000	0.883121	
concavity_mean	0.521984	0.883121	1.000000	
concave_points_mean	0.553695	0.831135	0.921391	
symmetry_mean	0.557775	0.602641	0.500667	
fractal_dimension_mean	0.584792	0.565369	0.336783	
radius_se	0.301467	0.497473	0.631925	
texture_se	0.068406	0.046205	0.076218	
perimeter_se	0.296092	0.548905	0.660391	
area_se	0.246552	0.455653	0.617427	
smoothness_se	0.332375	0.135299	0.098564	
compactness_se	0.318943	0.738722	0.670279	
concavity_se	0.248396	0.570517	0.691270	
concave_points_se	0.380676	0.642262	0.683260	
symmetry_se	0.200774	0.229977	0.178009	
fractal_dimension_se	0.283607	0.507318	0.449301	

radius_worst	0.213120	0.535315	0.688236
texture_worst	0.036072	0.248133	0.299879
perimeter_worst	0.238853	0.590210	0.729565
area_worst	0.206718	0.509604	0.675987
smoothness_worst	0.805324	0.565541	0.448822
compactness_worst	0.472468	0.865809	0.754968
concavity_worst	0.434926	0.816275	0.884103
concave_points_worst	0.503053	0.815573	0.861323
symmetry_worst	0.394309	0.510223	0.409464
fractal_dimension_worst	0.499316	0.687382	0.514930

	concave_points_mean	symmetry_mean	\
radius_mean	0.822529	0.147741	
texture_mean	0.293464	0.071401	
perimeter_mean	0.850977	0.183027	
area_mean	0.823269	0.151293	
smoothness_mean	0.553695	0.557775	
compactness_mean	0.831135	0.602641	
concavity_mean	0.921391	0.500667	
concave_points_mean	1.000000	0.462497	
symmetry_mean	0.462497	1.000000	
fractal_dimension_mean	0.166917	0.479921	
radius_se	0.698050	0.303379	
texture_se	0.021480	0.128053	
perimeter_se	0.710650	0.313893	
area_se	0.690299	0.223970	
smoothness_se	0.027653	0.187321	
compactness_se	0.490424	0.421659	
concavity_se	0.439167	0.342627	
concave_points_se	0.615634	0.393298	
symmetry_se	0.095351	0.449137	
fractal_dimension_se	0.257584	0.331786	
radius_worst	0.830318	0.185728	
texture_worst	0.292752	0.090651	
perimeter_worst	0.855923	0.219169	
area_worst	0.809630	0.177193	
smoothness_worst	0.452753	0.426675	
compactness_worst	0.667454	0.473200	
concavity_worst	0.752399	0.433721	
concave_points_worst	0.910155	0.430297	
symmetry_worst	0.375744	0.699826	
fractal_dimension_worst	0.368661	0.438413	

	fractal_dimension_mean	radius_se	texture_se	\
radius_mean	-0.311631	0.679090	-0.097317	
texture_mean	-0.076437	0.275869	0.386358	
perimeter_mean	-0.261477	0.691765	-0.086761	

area_mean	-0.283110	0.732562	-0.066280
smoothness_mean	0.584792	0.301467	0.068406
compactness_mean	0.565369	0.497473	0.046205
concavity_mean	0.336783	0.631925	0.076218
concave_points_mean	0.166917	0.698050	0.021480
symmetry_mean	0.479921	0.303379	0.128053
fractal_dimension_mean	1.000000	0.000111	0.164174
radius_se	0.000111	1.000000	0.213247
texture_se	0.164174	0.213247	1.000000
perimeter_se	0.039830	0.972794	0.223171
area_se	-0.090170	0.951830	0.111567
smoothness_se	0.401964	0.164514	0.397243
compactness_se	0.559837	0.356065	0.231700
concavity_se	0.446630	0.332358	0.194998
concave_points_se	0.341198	0.513346	0.230283
symmetry_se	0.345007	0.240567	0.411621
fractal_dimension_se	0.688132	0.227754	0.279723
radius_worst	-0.253691	0.715065	-0.111690
texture_worst	-0.051269	0.194799	0.409003
perimeter_worst	-0.205151	0.719684	-0.102242
area_worst	-0.231854	0.751548	-0.083195
smoothness_worst	0.504942	0.141919	-0.073658
compactness_worst	0.458798	0.287103	-0.092439
concavity_worst	0.346234	0.380585	-0.068956
concave_points_worst	0.175325	0.531062	-0.119638
symmetry_worst	0.334019	0.094543	-0.128215
fractal_dimension_worst	0.767297	0.049559	-0.045655

	perimeter_se	area_se	smoothness_se	\
radius_mean	0.674172	0.735864	-0.222600	
texture_mean	0.281673	0.259845	0.006614	
perimeter_mean	0.693135	0.744983	-0.202694	
area_mean	0.726628	0.800086	-0.166777	
smoothness_mean	0.296092	0.246552	0.332375	
compactness_mean	0.548905	0.455653	0.135299	
concavity_mean	0.660391	0.617427	0.098564	
concave_points_mean	0.710650	0.690299	0.027653	
symmetry_mean	0.313893	0.223970	0.187321	
fractal_dimension_mean	0.039830	-0.090170	0.401964	
radius_se	0.972794	0.951830	0.164514	
texture_se	0.223171	0.111567	0.397243	
perimeter_se	1.000000	0.937655	0.151075	
area_se	0.937655	1.000000	0.075150	
smoothness_se	0.151075	0.075150	1.000000	
compactness_se	0.416322	0.284840	0.336696	
concavity_se	0.362482	0.270895	0.268685	
concave_points_se	0.556264	0.415730	0.328429	

symmetry_se	0.266487	0.134109	0.413506
fractal_dimension_se	0.244143	0.127071	0.427374
radius_worst	0.697201	0.757373	-0.230691
texture_worst	0.200371	0.196497	-0.074743
perimeter_worst	0.721031	0.761213	-0.217304
area_worst	0.730713	0.811408	-0.182195
smoothness_worst	0.130054	0.125389	0.314457
compactness_worst	0.341919	0.283257	-0.055558
concavity_worst	0.418899	0.385100	-0.058298
concave_points_worst	0.554897	0.538166	-0.102007
symmetry_worst	0.109930	0.074126	-0.107342
fractal_dimension_worst	0.085433	0.017539	0.101480

	compactness_se	concavity_se	concave_points_se	\
radius_mean	0.206000	0.194204	0.376169	
texture_mean	0.191975	0.143293	0.163851	
perimeter_mean	0.250744	0.228082	0.407217	
area_mean	0.212583	0.207660	0.372320	
smoothness_mean	0.318943	0.248396	0.380676	
compactness_mean	0.738722	0.570517	0.642262	
concavity_mean	0.670279	0.691270	0.683260	
concave_points_mean	0.490424	0.439167	0.615634	
symmetry_mean	0.421659	0.342627	0.393298	
fractal_dimension_mean	0.559837	0.446630	0.341198	
radius_se	0.356065	0.332358	0.513346	
texture_se	0.231700	0.194998	0.230283	
perimeter_se	0.416322	0.362482	0.556264	
area_se	0.284840	0.270895	0.415730	
smoothness_se	0.336696	0.268685	0.328429	
compactness_se	1.000000	0.801268	0.744083	
concavity_se	0.801268	1.000000	0.771804	
concave_points_se	0.744083	0.771804	1.000000	
symmetry_se	0.394713	0.309429	0.312780	
fractal_dimension_se	0.803269	0.727372	0.611044	
radius_worst	0.204607	0.186904	0.358127	
texture_worst	0.143003	0.100241	0.086741	
perimeter_worst	0.260516	0.226680	0.394999	
area_worst	0.199371	0.188353	0.342271	
smoothness_worst	0.227394	0.168481	0.215351	
compactness_worst	0.678780	0.484858	0.452888	
concavity_worst	0.639147	0.662564	0.549592	
concave_points_worst	0.483208	0.440472	0.602450	
symmetry_worst	0.277878	0.197788	0.143116	
fractal_dimension_worst	0.590973	0.439329	0.310655	

	symmetry_se	fractal_dimension_se	radius_worst	\
radius_mean	-0.104321	-0.042641	0.969539	

texture_mean	0.009127	0.054458	0.352573
perimeter_mean	-0.081629	-0.005523	0.969476
area_mean	-0.072497	-0.019887	0.962746
smoothness_mean	0.200774	0.283607	0.213120
compactness_mean	0.229977	0.507318	0.535315
concavity_mean	0.178009	0.449301	0.688236
concave_points_mean	0.095351	0.257584	0.830318
symmetry_mean	0.449137	0.331786	0.185728
fractal_dimension_mean	0.345007	0.688132	-0.253691
radius_se	0.240567	0.227754	0.715065
texture_se	0.411621	0.279723	-0.111690
perimeter_se	0.266487	0.244143	0.697201
area_se	0.134109	0.127071	0.757373
smoothness_se	0.413506	0.427374	-0.230691
compactness_se	0.394713	0.803269	0.204607
concavity_se	0.309429	0.727372	0.186904
concave_points_se	0.312780	0.611044	0.358127
symmetry_se	1.000000	0.369078	-0.128121
fractal_dimension_se	0.369078	1.000000	-0.037488
radius_worst	-0.128121	-0.037488	1.000000
texture_worst	-0.077473	-0.003195	0.359921
perimeter_worst	-0.103753	-0.001000	0.993708
area_worst	-0.110343	-0.022736	0.984015
smoothness_worst	-0.012662	0.170568	0.216574
compactness_worst	0.060255	0.390159	0.475820
concavity_worst	0.037119	0.379975	0.573975
concave_points_worst	-0.030413	0.215204	0.787424
symmetry_worst	0.389402	0.111094	0.243529
fractal_dimension_worst	0.078079	0.591328	0.093492

	texture_worst	perimeter_worst	area_worst	\
radius_mean	0.297008	0.965137	0.941082	
texture_mean	0.912045	0.358040	0.343546	
perimeter_mean	0.303038	0.970387	0.941550	
area_mean	0.287489	0.959120	0.959213	
smoothness_mean	0.036072	0.238853	0.206718	
compactness_mean	0.248133	0.590210	0.509604	
concavity_mean	0.299879	0.729565	0.675987	
concave_points_mean	0.292752	0.855923	0.809630	
symmetry_mean	0.090651	0.219169	0.177193	
fractal_dimension_mean	-0.051269	-0.205151	-0.231854	
radius_se	0.194799	0.719684	0.751548	
texture_se	0.409003	-0.102242	-0.083195	
perimeter_se	0.200371	0.721031	0.730713	
area_se	0.196497	0.761213	0.811408	
smoothness_se	-0.074743	-0.217304	-0.182195	
compactness_se	0.143003	0.260516	0.199371	

concavity_se	0.100241	0.226680	0.188353
concave_points_se	0.086741	0.394999	0.342271
symmetry_se	-0.077473	-0.103753	-0.110343
fractal_dimension_se	-0.003195	-0.001000	-0.022736
radius_worst	0.359921	0.993708	0.984015
texture_worst	1.000000	0.365098	0.345842
perimeter_worst	0.365098	1.000000	0.977578
area_worst	0.345842	0.977578	1.000000
smoothness_worst	0.225429	0.236775	0.209145
compactness_worst	0.360832	0.529408	0.438296
concavity_worst	0.368366	0.618344	0.543331
concave_points_worst	0.359755	0.816322	0.747419
symmetry_worst	0.233027	0.269493	0.209146
fractal_dimension_worst	0.219122	0.138957	0.079647

	smoothness_worst	compactness_worst	concavity_worst	\
radius_mean	0.119616	0.413463	0.526911	
texture_mean	0.077503	0.277830	0.301025	
perimeter_mean	0.150549	0.455774	0.563879	
area_mean	0.123523	0.390410	0.512606	
smoothness_mean	0.805324	0.472468	0.434926	
compactness_mean	0.565541	0.865809	0.816275	
concavity_mean	0.448822	0.754968	0.884103	
concave_points_mean	0.452753	0.667454	0.752399	
symmetry_mean	0.426675	0.473200	0.433721	
fractal_dimension_mean	0.504942	0.458798	0.346234	
radius_se	0.141919	0.287103	0.380585	
texture_se	-0.073658	-0.092439	-0.068956	
perimeter_se	0.130054	0.341919	0.418899	
area_se	0.125389	0.283257	0.385100	
smoothness_se	0.314457	-0.055558	-0.058298	
compactness_se	0.227394	0.678780	0.639147	
concavity_se	0.168481	0.484858	0.662564	
concave_points_se	0.215351	0.452888	0.549592	
symmetry_se	-0.012662	0.060255	0.037119	
fractal_dimension_se	0.170568	0.390159	0.379975	
radius_worst	0.216574	0.475820	0.573975	
texture_worst	0.225429	0.360832	0.368366	
perimeter_worst	0.236775	0.529408	0.618344	
area_worst	0.209145	0.438296	0.543331	
smoothness_worst	1.000000	0.568187	0.518523	
compactness_worst	0.568187	1.000000	0.892261	
concavity_worst	0.518523	0.892261	1.000000	
concave_points_worst	0.547691	0.801080	0.855434	
symmetry_worst	0.493838	0.614441	0.532520	
fractal_dimension_worst	0.617624	0.810455	0.686511	

	concave_points_worst	symmetry_worst	\
radius_mean	0.744214	0.163953	
texture_mean	0.295316	0.105008	
perimeter_mean	0.771241	0.189115	
area_mean	0.722017	0.143570	
smoothness_mean	0.503053	0.394309	
compactness_mean	0.815573	0.510223	
concavity_mean	0.861323	0.409464	
concave_points_mean	0.910155	0.375744	
symmetry_mean	0.430297	0.699826	
fractal_dimension_mean	0.175325	0.334019	
radius_se	0.531062	0.094543	
texture_se	-0.119638	-0.128215	
perimeter_se	0.554897	0.109930	
area_se	0.538166	0.074126	
smoothness_se	-0.102007	-0.107342	
compactness_se	0.483208	0.277878	
concavity_se	0.440472	0.197788	
concave_points_se	0.602450	0.143116	
symmetry_se	-0.030413	0.389402	
fractal_dimension_se	0.215204	0.111094	
radius_worst	0.787424	0.243529	
texture_worst	0.359755	0.233027	
perimeter_worst	0.816322	0.269493	
area_worst	0.747419	0.209146	
smoothness_worst	0.547691	0.493838	
compactness_worst	0.801080	0.614441	
concavity_worst	0.855434	0.532520	
concave_points_worst	1.000000	0.502528	
symmetry_worst	0.502528	1.000000	
fractal_dimension_worst	0.511114	0.537848	

	fractal_dimension_worst
radius_mean	0.007066
texture_mean	0.119205
perimeter_mean	0.051019
area_mean	0.003738
smoothness_mean	0.499316
compactness_mean	0.687382
concavity_mean	0.514930
concave_points_mean	0.368661
symmetry_mean	0.438413
fractal_dimension_mean	0.767297
radius_se	0.049559
texture_se	-0.045655
perimeter_se	0.085433
area_se	0.017539

```
smoothness_se                   0.101480
compactness_se                  0.590973
concavity_se                    0.439329
concave_points_se               0.310655
symmetry_se                     0.078079
fractal_dimension_se            0.591328
radius_worst                    0.093492
texture_worst                   0.219122
perimeter_worst                 0.138957
area_worst                      0.079647
smoothness_worst                0.617624
compactness_worst               0.810455
concavity_worst                 0.686511
concave_points_worst            0.511114
symmetry_worst                  0.537848
fractal_dimension_worst         1.000000
```

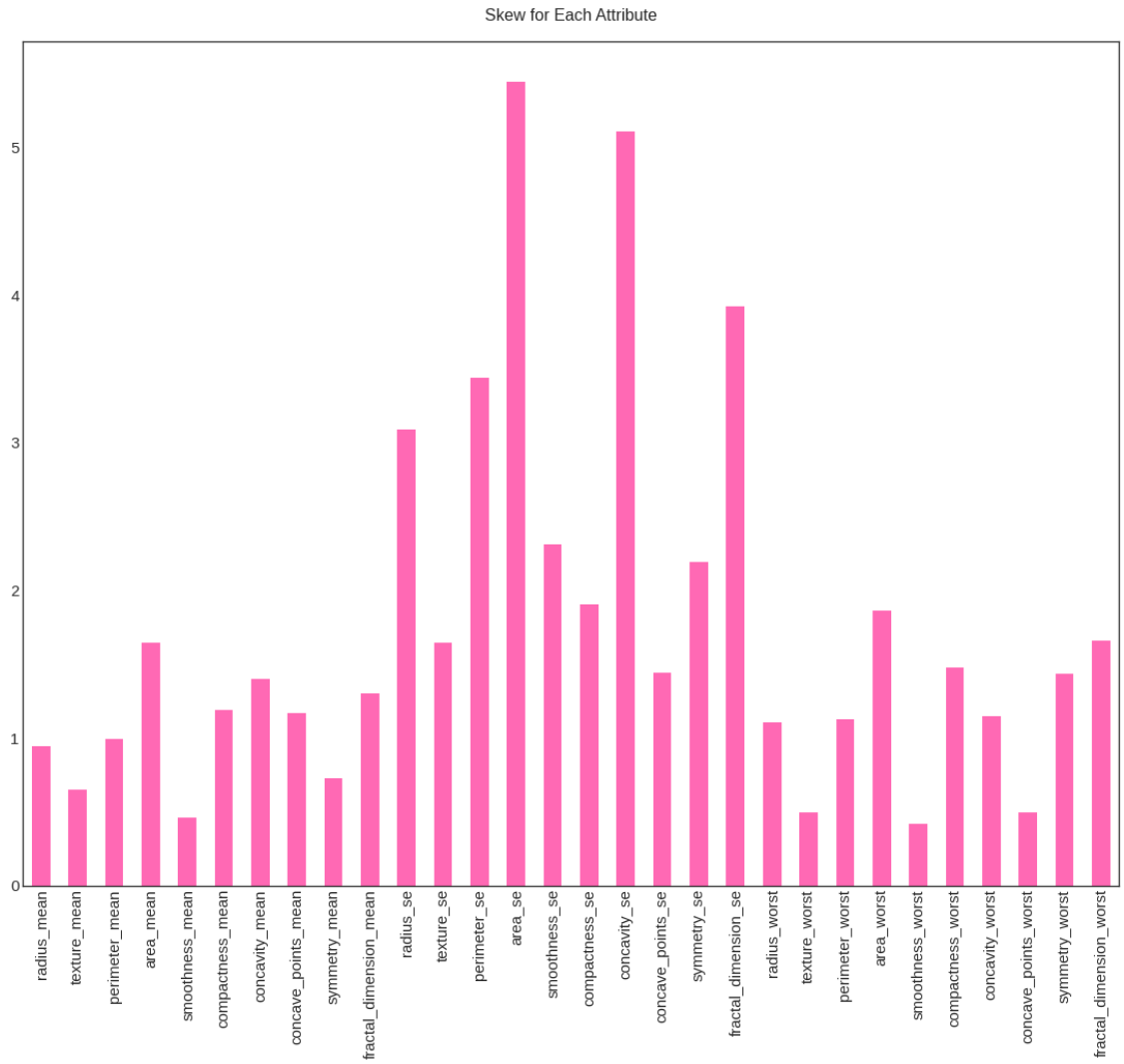
[6]:
```
pd.set_option('display.max_rows', 1000)

sorted_mat = dataset[features_pearson].corr(method='pearson').unstack().
 ↪sort_values(ascending=False)

# Print the 10 strongest positive pairwise correlations after self-correlations.
print('* Strongest Positive Correlations *\n', sorted_mat[31:].head(10))

# Print the 10 weakest pairwise correlations.
print('\n* Closest to Zero Correlations *\n', sorted_mat[:807].tail(10))

# Print the 10 strongest negative pairwise correlations.
print('\n* Strongest Negative Correlations *\n', sorted_mat.tail(10))
```

```
* Strongest Positive Correlations *
 perimeter_mean   radius_mean         0.997855
radius_worst     perimeter_worst     0.993708
perimeter_worst  radius_worst        0.993708
radius_mean      area_mean           0.987357
area_mean        radius_mean         0.987357
                 perimeter_mean      0.986507
perimeter_mean   area_mean           0.986507
area_worst       radius_worst        0.984015
radius_worst     area_worst          0.984015
perimeter_worst  area_worst          0.977578
dtype: float64

* Closest to Zero Correlations *
 smoothness_se            texture_mean              0.006614
area_mean                fractal_dimension_worst   0.003738
```

```
fractal_dimension_worst  area_mean                 0.003738
radius_se                fractal_dimension_mean    0.000111
fractal_dimension_mean   radius_se                 0.000111
perimeter_worst          fractal_dimension_se     -0.001000
fractal_dimension_se     perimeter_worst          -0.001000
texture_worst            fractal_dimension_se     -0.003195
fractal_dimension_se     texture_worst            -0.003195
                         perimeter_mean           -0.005523
dtype: float64

* Strongest Negative Correlations *
 fractal_dimension_mean  area_worst               -0.231854
area_worst               fractal_dimension_mean   -0.231854
radius_worst             fractal_dimension_mean   -0.253691
fractal_dimension_mean   radius_worst             -0.253691
                         perimeter_mean           -0.261477
perimeter_mean           fractal_dimension_mean   -0.261477
area_mean                fractal_dimension_mean   -0.283110
fractal_dimension_mean   area_mean                -0.283110
radius_mean              fractal_dimension_mean   -0.311631
fractal_dimension_mean   radius_mean              -0.311631
dtype: float64
```

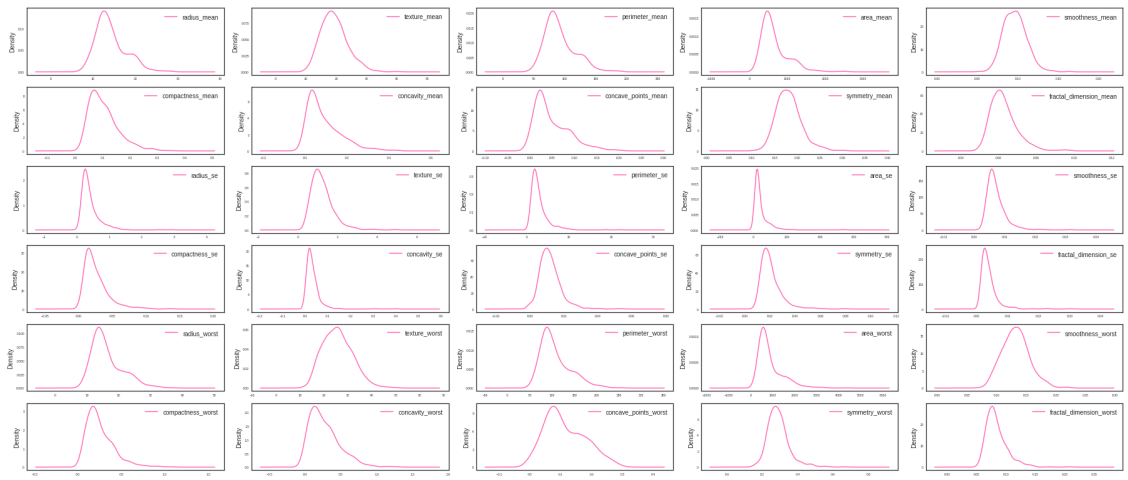
Skew for Each Attribute

[7]:
```
features = list(dataset.columns[1:32])
s = dataset[features].skew()
plt.figure(figsize=(19, 15))
s.plot(kind = 'bar', fontsize=15, color='hotpink')
plt.title('Skew for Each Attribute', fontsize=16, pad=20);
```



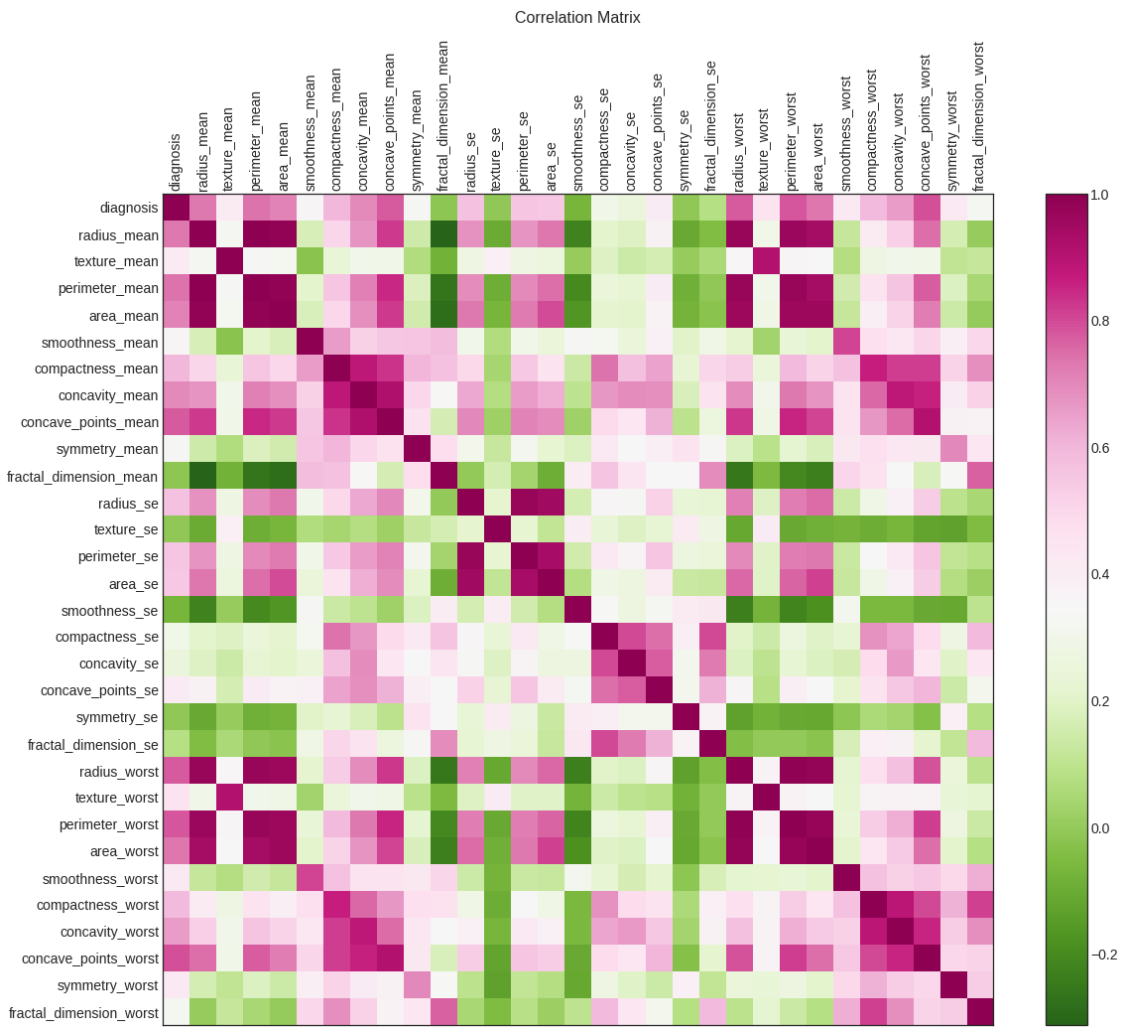
Univariate Density Plot

[8]:
```
dataset.select_dtypes('float').plot(kind='kde', subplots=True, layout=(20,5),␣
↪figsize=(25,35),
             sharex=False, sharey=False, fontsize=5, color='hotpink')
plt.tight_layout()
plt.show()
```



Correlation Matrix Plot

[9]:
```
f = plt.figure(figsize=(19, 15))
features = list(dataset.columns[0:32])
plt.matshow(dataset[features].corr(), fignum=f.number, cmap='PiYG_r')
plt.xticks(range(dataset[features].select_dtypes(['number']).shape[1]),␣
 ↪dataset[features].select_dtypes(['number']).columns, fontsize=14,␣
 ↪rotation=90)
plt.yticks(range(dataset[features].select_dtypes(['number']).shape[1]),␣
 ↪dataset[features].select_dtypes(['number']).columns, fontsize=14)
cb = plt.colorbar()
cb.ax.tick_params(labelsize=14)
plt.title('Correlation Matrix', fontsize=16, pad=130);
```



4 Listing 7

Rescaling Data, Standardize Data, Normalize Data, Binarize Data. Analysis

The unit of measurement for each feature was not provided in the data dictionary documentation. Features such as radius have a different unit and scale than those such as texture or symmetry. Rescaling allows for all values to fall between the given interval, [0,1]. For a given feature, the new value is proportional to the original value's place inside the spread of [0, max(feature)]. For example, the texture_mean of the first value (10.38) lies closer to the original minimum (9.71) than the 25th percentile (16.17). Rescaling transforms the value from 10.38 to 0.023: (10.38 - 9.71) / (39.28 - 9.71) = 0.023

Standardization (Z-score normalization) of the data assumes Gaussian distributions of real-valued features. By standardizing to a mean of 0 and a standard deviation of 1, better performance should be seen with linear algorithms such as logistic regression and linear discriminant analysis.

The minimum and maximum values are highly influenced by the presence of outliers, which may increase the overall spread. Minimum and maximum values for each feature vary, dependent on the original spread of each feature.

Normalization does not assume Gaussian distribution, and lends itself to algorithms that don't make this assumption, such as k-nearest neighbors and artificial neural networks. By having the bound at a fixed range, 0 to 1, the smaller standard deviations lead to sensitivity to outliers.

Binarization transforms the values to 0 or 1, dependent on whether each value falls below or above a given threshold. The features of this dataset all have a theoretical minimum of 0, but differing maximums. Given the differences in spread, rather than fix the threshold to a constant, we vary it based on the mean of each feature. This can be achieved by providing a list of the 30 feature means as the binarizer threshold, or by fitting and transforming on a standardized dataset and setting the threshold to the standardized mean of 0.

[10]:
```
# Set precision.
np.set_printoptions(precision=3)
```

[11]:
```
# Separate out target variable and print original values for first two rows.
X = dataset.values[:, 1:]
Y = dataset.values[:, 0]
print(X[0:2, :])
```

```
[[1.799e+01 1.038e+01 1.228e+02 1.001e+03 1.184e-01 2.776e-01 3.001e-01
  1.471e-01 2.419e-01 7.871e-02 1.095e+00 9.053e-01 8.589e+00 1.534e+02
  6.399e-03 4.904e-02 5.373e-02 1.587e-02 3.003e-02 6.193e-03 2.538e+01
  1.733e+01 1.846e+02 2.019e+03 1.622e-01 6.656e-01 7.119e-01 2.654e-01
  4.601e-01 1.189e-01]
 [2.057e+01 1.777e+01 1.329e+02 1.326e+03 8.474e-02 7.864e-02 8.690e-02
  7.017e-02 1.812e-01 5.667e-02 5.435e-01 7.339e-01 3.398e+00 7.408e+01
  5.225e-03 1.308e-02 1.860e-02 1.340e-02 1.389e-02 3.532e-03 2.499e+01
  2.341e+01 1.588e+02 1.956e+03 1.238e-01 1.866e-01 2.416e-01 1.860e-01
  2.750e-01 8.902e-02]]
```

a) Rescaling

[12]:
```
scaler_rescaled = MinMaxScaler(feature_range=(0, 1))
X_rescaled = scaler_rescaled.fit_transform(X)
print(X_rescaled[0:2, :])
```

```
[[0.521 0.023 0.546 0.364 0.594 0.792 0.703 0.731 0.686 0.606 0.356 0.12
  0.369 0.274 0.159 0.351 0.136 0.301 0.312 0.183 0.621 0.142 0.668 0.451
  0.601 0.619 0.569 0.912 0.598 0.419]
 [0.643 0.273 0.616 0.502 0.29  0.182 0.204 0.349 0.38  0.141 0.156 0.083
  0.124 0.126 0.119 0.081 0.047 0.254 0.085 0.091 0.607 0.304 0.54  0.435
  0.348 0.155 0.193 0.639 0.234 0.223]]
```

b) Standardization

[13]:
```
scaler_standardized = StandardScaler()
X_standardized = scaler_standardized.fit_transform(X)
print(X_standardized[0:2, :])
```

```
[[ 1.097e+00 -2.073e+00  1.270e+00  9.844e-01  1.568e+00  3.284e+00
   2.653e+00  2.532e+00  2.218e+00  2.256e+00  2.490e+00 -5.653e-01
   2.833e+00  2.488e+00 -2.140e-01  1.317e+00  7.240e-01  6.608e-01
   1.149e+00  9.071e-01  1.887e+00 -1.359e+00  2.304e+00  2.001e+00
   1.308e+00  2.617e+00  2.110e+00  2.296e+00  2.751e+00  1.937e+00]
 [ 1.830e+00 -3.536e-01  1.686e+00  1.909e+00 -8.270e-01 -4.871e-01
  -2.385e-02  5.481e-01  1.392e-03 -8.687e-01  4.993e-01 -8.762e-01
   2.633e-01  7.424e-01 -6.054e-01 -6.929e-01 -4.408e-01  2.602e-01
  -8.055e-01 -9.944e-02  1.806e+00 -3.692e-01  1.535e+00  1.890e+00
  -3.756e-01 -4.304e-01 -1.467e-01  1.087e+00 -2.439e-01  2.812e-01]]
```

c) Normalization

[14]:
```
scaler_normalized = Normalizer()
X_normalized = scaler_normalized.fit_transform(X)
print(X_normalized[0:2, :])
```

```
[[7.925e-03 4.573e-03 5.410e-02 4.410e-01 5.216e-05 1.223e-04 1.322e-04
  6.480e-05 1.066e-04 3.468e-05 4.824e-04 3.988e-04 3.784e-03 6.758e-02
  2.819e-06 2.160e-05 2.367e-05 6.991e-06 1.323e-05 2.728e-06 1.118e-02
  7.635e-03 8.132e-02 8.895e-01 7.146e-05 2.932e-04 3.136e-04 1.169e-04
  2.027e-04 5.238e-05]
 [8.666e-03 7.486e-03 5.599e-02 5.586e-01 3.570e-05 3.313e-05 3.661e-05
  2.956e-05 7.634e-05 2.387e-05 2.290e-04 3.092e-04 1.432e-03 3.121e-02
  2.201e-06 5.510e-06 7.836e-06 5.645e-06 5.852e-06 1.488e-06 1.053e-02
  9.862e-03 6.690e-02 8.240e-01 5.215e-05 7.861e-05 1.018e-04 7.836e-05
  1.159e-04 3.750e-05]]
```

d) Binarization

[15]:
```
# Print the 30 feature means.
print(X.mean(axis=0))

# Binarize X with the means as thresholds.
binarizer = Binarizer(threshold=X.mean(axis=0)).fit(X)
X_binarized = binarizer.transform(X)
print(X_binarized[0:2, :])
```

```
[1.413e+01 1.929e+01 9.197e+01 6.549e+02 9.636e-02 1.043e-01 8.880e-02
 4.892e-02 1.812e-01 6.280e-02 4.052e-01 1.217e+00 2.866e+00 4.034e+01
 7.041e-03 2.548e-02 3.189e-02 1.180e-02 2.054e-02 3.795e-03 1.627e+01
 2.568e+01 1.073e+02 8.806e+02 1.324e-01 2.543e-01 2.722e-01 1.146e-01
 2.901e-01 8.395e-02]
[[1. 0. 1. 1. 1. 1. 1. 1. 1. 1. 1. 0. 1. 1. 0. 1. 1. 1. 1. 1. 1. 0. 1. 1.
  1. 1. 1. 1. 1. 1.]
```

```
 [1. 0. 1. 1. 0. 0. 0. 1. 1. 0. 1. 0. 1. 1. 0. 0. 0. 1. 0. 0. 1. 0. 1. 1.
  0. 0. 0. 1. 0. 1.]]
```

[16]:
```
# Alternatively, binarize the standarized X with the threshold at the standard␣
 ↪mean of 0.
binarizer = Binarizer(threshold=0).fit(X_standardized)
X_binarized = binarizer.transform(X_standardized)
print(X_binarized[0:2, :])
```

```
[[1. 0. 1. 1. 1. 1. 1. 1. 1. 1. 1. 0. 1. 1. 0. 1. 1. 1. 1. 1. 1. 0. 1. 1.
  1. 1. 1. 1. 1. 1.]
 [1. 0. 1. 1. 0. 0. 0. 1. 1. 0. 1. 0. 1. 1. 0. 0. 0. 1. 0. 0. 1. 0. 1. 1.
  0. 0. 0. 1. 0. 1.]]
```