Other Dosage Forms

```
0.00
In [1]:
        SELECT
        ag_custom_case_number__c,
        prod.name,
        dose_form.name,
        ROW NUMBER() OVER(PARTITION BY ca.ag custom case number c, pic.name ORDER BY contents
        pic.name as issue type,
        convert from(decode(contentver.versiondata, 'base64'),'UTF8') AS notes
        FROM
        bct schema."CASE" ca inner join bct schema."CASE HISTORY" casehistory on ca.id = case
        inner join bct_schema."CONTENTDOCUMENTLINK" contdoclink on ca.parentid = contdoclink.1
        inner join bct_schema."CONTENTVERSION" contentver on contdoclink.contentdocumentid = 
        inner join bct_schema."AG_PRODUCT__C" prod on ca.ag_product__c = prod.id
        inner join bct_schema."AG_CASE_PRODUCT__C" cprod on ca.parentid = cprod.ag_case__c
        inner join bct schema."AG DOSAGE FORM C" dose form on cprod.ag dosage form c = dose
        inner join bct schema."AG PCM ISSUE C" pi on ca.ag custom case number c = pi.ag pcm
        inner join bct schema."AG PCM ISSUE CODE C" pic on pic.id = pi.ag as reported code
        inner join bct_schema."AG_PCM_ISSUE_CODE_FAMILY__C" picf on pi.ag_cause_code_family__c
        dose form.name IN ('Solution for injection in pre-filled pen','Vial - liquid','Vial -
        ca.ag intake channel type c IS NOT NULL and
        casehistory.field = 'Status' and
        casehistory.newvalue = 'Intake Complete' and
        contentver.src_createddate <= casehistory.src_createddate;</pre>
        0.00
```

'\nSELECT\nag_custom_case_number__c,\nprod.name,\ndose_form.name,\nROW_NUMBER() OVER Out[1]: (PARTITION BY ca.ag custom case number c, pic.name ORDER BY contentver.src createdda te DESC) AS notes_rank,\npic.name as issue_type,\nconvert_from(decode(contentver.vers iondata, \'base64\'),\'UTF8\') AS notes\nFROM \nbct_schema."CASE" ca inner join bct_s chema."CASE HISTORY" casehistory on ca.id = casehistory.caseid \ninner join bct sche ma."CONTENTDOCUMENTLINK" contdoclink on ca.parentid = contdoclink.linkedentityid\ninn er join bct schema. "CONTENTVERSION" contentver on contdoclink.contentdocumentid = con tentver.contentdocumentid\ninner join bct schema."AG PRODUCT C" prod on ca.ag produc t__c = prod.id\ninner join bct_schema."AG_CASE_PRODUCT__C" cprod on ca.parentid = cpr od.ag_case__c\ninner join bct_schema."AG_DOSAGE_FORM__C" dose_form on cprod.ag_dosage form c = dose form.id\ninner join bct schema."AG PCM ISSUE C" pi on ca.ag custom c ase number c = pi.ag pcm sub case number apex c\ninner join bct schema."AG PCM ISS UE CODE C" pic on pic.id = pi.ag as reported code c\ninner join bct schema."AG PCM ISSUE_CODE_FAMILY__C" picf on pi.ag_cause_code_family__c = picf.id\nWHERE\ndose_form. name IN (\'Solution for injection in pre-filled pen\',\'Vial - liquid\',\'Vial - lyop hilized\',\'Software based device\',\'Tablet\') and \nca.ag intake channel type c IS NOT NULL and\ncasehistory.field = \'Status\' and \ncasehistory.newvalue = \'Intake Co mplete\' and \ncontentver.src createddate <= casehistory.src createddate; \n'</pre>

```
In [2]: import sys
    print(sys.executable)
    print(sys.version)
    print(sys.version_info)

    C:\Users\gmodi\Anaconda3\envs\FastText\python.exe
    3.7.13 (default, Mar 28 2022, 08:03:21) [MSC v.1916 64 bit (AMD64)]
    sys.version_info(major=3, minor=7, micro=13, releaselevel='final', serial=0)

In [3]: import numpy as np
    import pandas as pd
```

```
import texthero as hero
from texthero import stopwords
from texthero import preprocessing
from texthero import visualization
from texthero import representation

from bs4 import BeautifulSoup

import fasttext
from gensim.utils import simple_preprocess
from gensim.parsing.preprocessing import remove_stopwords

from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report
from sklearn.metrics import roc_auc_score
from sklearn import preprocessing
```

In [10]: masterData = pd.read_csv("C:/Users/gmodi/Downloads/data-1666644445166.csv")
#masterData = pd.read_csv("C:/Users/gmodi/Downloads/data-1671237323737.csv")
#masterData["issue_type"] = masterData["issue_type"].str.replace(' ', '_').replace('/masterData["issue_type"] = masterData["issue_type"].str.replace(r'[^0-9a-zA-Z:,]+', ']
masterData["len"] = masterData["notes"].apply(len)

C:\Users\gmodi\Anaconda3\envs\FastText\lib\site-packages\ipykernel_launcher.py:4: Fut
ureWarning: The default value of regex will change from True to False in a future ver
sion.

after removing the cwd from sys.path.

In [11]: masterData.head(5)

| Out[11]: | | ag_custom_case_numberc | product | name | notes_rank | issue_type | | |
|----------|-----|--|---------|--|------------|------------------------------------|------------------|--|
| | 0 | 19-0000093-PC-01 | Aranesp | Solution for injection in pre- filled pen | 1 | autoinjector_activation_difficulty | / reported \(\) | |
| | 1 | 19-0000093-PC-01 | Aranesp | Solution for injection in pre- filled pen | 2 | autoinjector_activation_difficulty | br>Prod | |
| | 2 | 19-0000745-PC-01 | Enbrel | Solution for injection in pre- filled pen | 1 | drug_injection | Date | |
| | 3 | 19-0000745-PC-01 | Enbrel | Solution for injection in pre- filled pen | 2 | drug_injection | | |
| | 4 | 19-0000745-PC-01 | Enbrel | Solution for injection in pre- filled pen | 3 | drug_injection | Tr numbs | |
| 4 | | | | | | | • | |
| In [12]: | # F | unctions | | | | | | |
| In [13]: | | <pre>def clean_notes(text): import re soup = BeautifulSoup(text, 'html.parser') list1 = [item.get_text() for item in list(soup.children)] list2 = [i for i in list1 if len(i) == max([len(i) for i in list1])] list3 = [re.sub('[^a-zA-Z:]+', ' ', _) for _ in list2] return list3[0] def multiclass_roc_auc_score(y_test, y_pred, average="macro"): lb = preprocessing.LabelBinarizer() lb.fit(y_test) y_test = lb.transform(y_test) y_pred = lb.transform(y_pred) return roc_auc_score(y_test, y_pred, average=average)</pre> | | | | | | |
| In [14]: | def | <pre>def normalize(s): """ Given a text, cleans and normalizes it. Feel free to add your own stuff. """</pre> | | | | | | |

```
s = s.lower()
# Replace ips
s = re.sub(r'\d{1,3}\.\d{1,3}\.\d{1,3}\.\d{1,3}\.\d{1,3}\', '__ip__', s)
# Isolate punctuation
s = re.sub(r'([.\(\)\!?\-\\\/\,])', r' \1 ', s)
# Remove some special characters
s = re.sub(r'([\;:\\]\•«\n])', '', s)
# Replace numbers and symbols with language
s = s.replace('&', ' and ')
s = s.replace('@', ' at ')
s = s.replace('0', ' zero ')
s = s.replace('1', ' one ')
s = s.replace('2', ' two ')
s = s.replace('3', ' three ')
s = s.replace('4', ' four ')
s = s.replace('5', ' five ')
s = s.replace('6', ' six ')
s = s.replace('8', ' eight ')
s = s.replace('9', ' nine ')
return s
```

Solution for injection in pre-filled pen

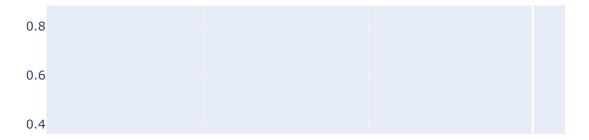
```
In [15]: data = masterData.query(" name == 'Solution for injection in pre-filled pen' ").copy()
    valueCount = data["issue_type"].value_counts(normalize=True).to_frame().cumsum()*100
    data = data[data["issue_type"].isin(valueCount.index.tolist()[0:9])]
    data = data[["notes","issue_type","len"]]

    data['notes']=data['notes'].apply(lambda cw : clean_notes(cw))
    data = data.query(" len > 500 ")

    data["notes"] = data["notes"].apply(lambda x: ' '.join(simple_preprocess(x, min_len=4, data["notes"] = data["notes"].apply(lambda x: remove_stopwords(''.join(x)))

In [16]: data["pca"] = (data["notes"].pipe(representation.tfidf, max_features=100).pipe(representation.scatterplot(data, col="pca", color="issue_type", title="PCA issue_type")
```

PCA issue_type





```
data["labeled_notes"] = data["issue_type"].apply(lambda x: '__label__' + x + " " ) + c
In [17]:
         x_train,x_test,y_train,y_test = train_test_split(data[["labeled_notes","issue_type"]],
          x train.to csv("C:/Users/gmodi/Downloads/x train.csv",index=False,header=False)
         x_test.to_csv("C:/Users/gmodi/Downloads/x_test.csv",index=False,header=False)
         model = fasttext.train supervised(input="C:/Users/gmodi/Downloads/x train.csv", wordNg
         model.test("C:/Users/gmodi/Downloads/x test.csv", k=3)
         (1636, 0.33027709861450694, 0.9908312958435208)
Out[17]:
In [18]:
         # predict the data
         x_test["predicted"] = x_test["labeled_notes"].apply(lambda x: model.predict(x)[0][0]).
         #Create the confusion matrix
          print(classification_report(x_test["issue_type"], x_test["predicted"]))
          print(confusion_matrix(x_test["issue_type"], x_test["predicted"]))
         multiclass_roc_auc_score(x_test["issue_type"], x_test["predicted"])
```

```
precision
                                                            recall f1-score
                                                                                support
                  Before Activation resolved
                                                    1.00
                                                              0.50
                                                                         0.67
                                                                                     10
              activation_difficulty_resolved
                                                    0.98
                                                              0.95
                                                                         0.97
                                                                                    172
          autoinjector activation difficulty
                                                                                   1369
                                                    0.97
                                                              0.96
                                                                         0.96
               autoinjector_user_mishandling
                                                    0.05
                                                              0.60
                                                                         0.09
                                                                                      5
                carton cosmetic minor damage
                                                    1.00
                                                              0.43
                                                                         0.60
                                                                                      7
              carton_label_missing_incorrect
                                                              0.50
                                                                         0.67
                                                                                      2
                                                    1.00
                                                                         0.83
                                                                                     17
                           customer_feedback
                                                    1.00
                                                              0.71
                              drug injection
                                                                         0.84
                                                                                     52
                                                    0.93
                                                              0.77
                                        other
                                                    1.00
                                                              1.00
                                                                         1.00
                                                                                      2
                                                                         0.94
                                                                                   1636
                                     accuracy
                                                              0.71
                                                                         0.74
                                   macro avg
                                                    0.88
                                                                                   1636
                                weighted avg
                                                    0.97
                                                              0.94
                                                                         0.95
                                                                                   1636
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                         8
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                    0
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         0.8468878491916008
Out[18]:
         x_test["prediction"] = x_test["labeled_notes"].apply(lambda x: model.predict(x,3)).ast
In [19]:
          #x_test["prediction"] = x_test["prediction"].astype(str)
          #x test["prediction"] = x test["prediction"].str.replace(' label ','')
          for i in range(len(x test)):
              if x_test.issue_type.iloc[i] in x_test.prediction.iloc[i]: x_test.predicted.iloc[i
              else: x test.predicted.iloc[i] = 0
          #x test.to csv("C:/Users/qmodi/Downloads/x test results.csv")
          x test["predicted"].value counts(normalize=True)*100
               95.904645
Out[19]:
                4.095355
         Name: predicted, dtype: float64
          model.save model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText SIPFP.bin")
In [20]:
          model.save model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText SIPFP.ftz")
```

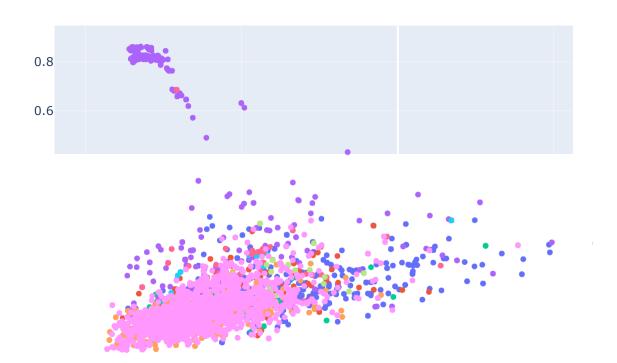
Vial - liquid

```
In [21]: #data = masterData.query(" name == 'Vial - liquid' ").copy()
    data = masterData
    valueCount = data["issue_type"].value_counts(normalize=True).to_frame().cumsum()*100
    data = data[data["issue_type"].isin(valueCount.index.tolist()[0:9])]
    data = data[["notes","issue_type","len"]]

data['notes']=data['notes'].apply(lambda cw : clean_notes(cw))
    data = data.query(" len > 500 ")
```

hero.scatterplot(data, col="pca", color="issue_type", title="PCA issue_type")

PCA issue type



```
In [23]: data["labeled_notes"] = data["issue_type"].apply(lambda x: '__label__' + x + " " ) + c
    x_train,x_test,y_train,y_test = train_test_split(data[["labeled_notes","issue_type"]],
    x_train.to_csv("C:/Users/gmodi/Downloads/x_train.csv",index=False,header=False)
    x_test.to_csv("C:/Users/gmodi/Downloads/x_test.csv",index=False,header=False)

model = fasttext.train_supervised(input="C:/Users/gmodi/Downloads/x_train.csv", wordNgmodel.test("C:/Users/gmodi/Downloads/x_test.csv",k=3)

Out[23]: (2341, 0.330485547486829, 0.9914566424604869)

In [24]: # predict the data
    x_test["predicted"] = x_test["labeled_notes"].apply(lambda x: model.predict(x)[0][0]).

#Create the confusion matrix
    confusion_matrix(x_test["issue_type"], x_test["predicted"])

print(classification_report(x_test["issue_type"], x_test["predicted"]))
```

```
print(confusion_matrix(x_test["issue_type"], x_test["predicted"]))
          multiclass_roc_auc_score(x_test["issue_type"], x_test["predicted"])
                                                   precision
                                                                recall f1-score
                                                                                    support
                  activation difficulty resolved
                                                        0.98
                                                                  0.90
                                                                             0.94
                                                                                        109
              autoinjector_activation_difficulty
                                                        0.95
                                                                  0.95
                                                                             0.95
                                                                                        921
                    carton cosmetic minor damage
                                                        0.99
                                                                  0.95
                                                                             0.97
                                                                                        380
                               customer feedback
                                                        0.90
                                                                  0.95
                                                                             0.92
                                                                                        515
                                  drug_injection
                                                        0.72
                                                                  0.84
                                                                             0.78
                                                                                         62
                                  drug_particles
                                                        0.67
                                                                  0.92
                                                                             0.78
                                                                                         61
          software_based_device_result_incorrect
                                                        0.95
                                                                  0.78
                                                                             0.86
                                                                                        151
           software based device technical issue
                                                        1.00
                                                                  0.71
                                                                             0.83
                                                                                         28
                  vial stopper damaged defective
                                                                  0.98
                                                                             0.99
                                                        0.99
                                                                                        114
                                        accuracy
                                                                             0.93
                                                                                       2341
                                                        0.91
                                                                  0.89
                                                                             0.89
                                                                                       2341
                                        macro avg
                                                                  0.93
                                    weighted avg
                                                        0.94
                                                                             0.93
                                                                                       2341
          [[ 98
                  9
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              2 875
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                             12
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                                               0]
              0
                  1
                      0
                          0
                                  0
                                       0
                                           0 112]]
                              1
         0.9385917238867003
Out[24]:
          x test["prediction"] = x test["labeled notes"].apply(lambda x: model.predict(x,3)).ast
In [25]:
          #x_test["prediction"] = x_test["prediction"].astype(str)
          #x_test["prediction"] = x_test["prediction"].str.replace('__label__','')
          x test["predicted"] = ""
          for i in range(len(x test)):
              if x_test.issue_type.iloc[i] in x_test.prediction.iloc[i]: x_test.predicted.iloc[i
              else: x test.predicted.iloc[i] = 0
          #x_test.to_csv("C:/Users/gmodi/Downloads/x_test_results.csv")
          x_test["predicted"].value_counts(normalize=True)*100
               98.077745
Out[25]:
                1.922255
         Name: predicted, dtype: float64
         model.save model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText Vial liquid.bin"
In [26]:
          model.save model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText Vial liquid.ftz"
```

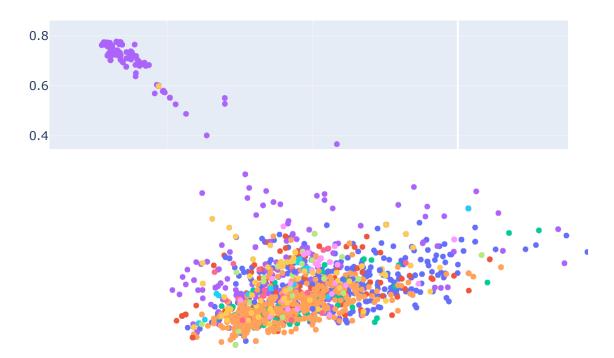
Vial - lyophilized

```
In [27]: #data = masterData.query(" name == 'Vial - Lyophilized' ").copy()
data = masterData
valueCount = data["issue_type"].value_counts(normalize=True).to_frame().cumsum()*100
data = data[data["issue_type"].isin(valueCount.index.tolist()[0:14])]
data = data[["notes","issue_type","len"]]
```

```
data['notes']=data['notes'].apply(lambda cw : clean_notes(cw))
                            data = data.query(" len > 200 ")
                           data["notes"] = data["notes"].apply(lambda x: ' '.join(simple_preprocess(x, min_len=4)
                            data["notes"] = data["notes"].apply(lambda x: remove stopwords(''.join(x)))
                           valueCount.head(20)
In [28]:
Out[28]:
                                                                                                                           issue_type
                                             autoinjector_activation_difficulty
                                                                                                                            33.503568
                                                                           customer_feedback
                                                                                                                            54.691285
                                               carton_cosmetic_minor_damage
                                                                                                                            68.840437
                                                                                                                           74.680809
                               software_based_device_result_incorrect
                                              vial_stopper_damaged_defective
                                                                                                                            78.511098
                                                      activation_difficulty_resolved
                                                                                                                            82.151509
                                                                                      drug_injection
                                                                                                                            84.718130
                                                                                       drug_particles
                                                                                                                            86.767498
                                 software_based_device_technical_issue
                                                                                                                            88.181759
                                                                                drug_appearance
                                                                                                                            89.170431
                                                                                                                            90.080534
                                                                                    needle_missing
                                       vial_plastic_cap_damaged_defective
                                                                                                                            90.689452
                                                                                                          other
                                                                                                                            91.291822
                                                                                                                            91.782885
                            interface_vial_adapter_leakage_breakage
                                          needle_blister_damaged_defective
                                                                                                                            92.254305
                                                                                 interface_needle
                                                                                                                            92.692988
                                                                                 drug_fill_volume
                                                                                                                            93.098933
                                        vial_kit_user_mishandling_difficulty
                                                                                                                            93.478688
                                                  autoinjector_user_mishandling
                                                                                                                            93.845348
                                                          Before_Activation_resolved
                                                                                                                            94.205461
                           data = data.query(" issue type not in ['customer feedback','other','To be determined']
In [29]:
In [30]:
                           data["pca"] = (data["notes"].pipe(representation.tfidf, max_features=100).pipe(representation.tfidf, max_f
                            hero.scatterplot(data, col="pca", color="issue_type", title="PCA issue_type")
```

```
localhost:8888/nbconvert/html/OtherDosageForms.ipynb?download=false
```

PCA issue_type



Out[32]:

Out[33]:

In [34]:

```
recall f1-score
                                                       precision
                                                                                         support
                    activation difficulty resolved
                                                            0.99
                                                                       0.93
                                                                                  0.96
                                                                                              147
               autoinjector_activation_difficulty
                                                            0.96
                                                                       0.96
                                                                                  0.96
                                                                                             1466
                      carton cosmetic minor damage
                                                            0.96
                                                                       0.97
                                                                                  0.97
                                                                                              565
                                    drug appearance
                                                            0.46
                                                                       0.60
                                                                                  0.53
                                                                                               43
                                     drug injection
                                                            0.64
                                                                       0.75
                                                                                  0.69
                                                                                              109
                                     drug_particles
                                                            0.95
                                                                       0.79
                                                                                  0.86
                                                                                              107
          interface_vial_adapter_leakage_breakage
                                                            0.68
                                                                       0.94
                                                                                  0.79
                                                                                               18
                                     needle missing
                                                            1.00
                                                                       1.00
                                                                                  1.00
                                                                                               35
           software based device result incorrect
                                                            0.92
                                                                       0.91
                                                                                  0.92
                                                                                              223
            software based device technical issue
                                                            0.81
                                                                       0.67
                                                                                  0.74
                                                                                               52
               vial_plastic_cap_damaged_defective
                                                            0.96
                                                                       0.96
                                                                                  0.96
                                                                                               27
                    vial_stopper_damaged_defective
                                                            1.00
                                                                       0.98
                                                                                  0.99
                                                                                              170
                                                                                  0.94
                                                                                             2962
                                            accuracy
                                                            0.86
                                                                       0.87
                                                                                  0.86
                                                                                             2962
                                           macro avg
                                                            0.94
                                                                       0.94
                                                                                  0.94
                                                                                             2962
                                       weighted avg
          [[ 137
                     9
                          0
                                0
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                                                                           0]
               0
                     3
                          1
                                4
                                     5
                                          0
                                                0
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                                                         204
                                                                6
                                                                           0]
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                          2
                                     0
                                                               35
                                                                      0
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                                                0
                                                     0
                                                          12
                                1
               0
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                                           0
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                                                           0
                                                                0
                                                                     26
                                                                           0]
               0
                     1
                          1
                                0
                                     0
                                           1
                                                0
                                                     0
                                                           0
                                                                0
                                                                      0
                                                                         167]]
          0.933262441260491
          x test["prediction"] = x test["labeled notes"].apply(lambda x: model.predict(x,3)).ast
In [33]:
          #x test["prediction"] = x test["prediction"].astype(str)
          #x_test["prediction"] = x_test["prediction"].str.replace('__label__','')
          x test["predicted"] = ""
          for i in range(len(x test)):
              if x test.issue type.iloc[i] in x test.prediction.iloc[i]: x test.predicted.iloc[i
              else: x_test.predicted.iloc[i] = 0
          #x_test.to_csv("C:/Users/gmodi/Downloads/x_test_results.csv")
          x test["predicted"].value counts(normalize=True)*100
               98.345712
                1.654288
          Name: predicted, dtype: float64
          model.words
```

```
['patient',
Out[34]:
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          model.labels
In [35]:
          ['__label__autoinjector_activation_difficulty',
Out[35]:
             _label__carton_cosmetic_minor_damage',
             _label__software_based_device_result_incorrect',
             label vial stopper damaged defective',
            __label__activation_difficulty_resolved',
              _label__drug_injection',
             label drug particles',
           '__label__software_based_device_technical_issue',
           '__label__needle_missing',
           '__label__drug_appearance',
             _label__vial_plastic_cap_damaged_defective',
           ' label interface vial adapter leakage breakage']
          model.wordNgrams
In [36]:
Out[36]:
```

```
model.get word vector('drug particles').shape
In [37]:
          (100,)
Out[37]:
In [38]:
          model.get nearest neighbors('overdose')
          [(0.9925181269645691, 'therapeutic'),
Out[38]:
           (0.9921038150787354, 'transmission'),
           (0.9917119741439819, 'intentional'),
           (0.9909931421279907, 'infectious'), (0.9909679889678955, 'occupational'),
           (0.990552544593811, 'abuse'),
           (0.9905371069908142, 'misuse'),
           (0.9901981353759766, 'findings'),
           (0.9888617992401123, 'benefit'),
           (0.9865464568138123, 'accidental')]
In [39]: model.get nearest neighbors('interface needle')
         [(0.0, 'aranesp'),
Out[39]:
           (0.0, '</s>'),
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           (0.0, 'product'),
           (0.0, 'thursdays'),
           (0.0, 'statement,software_based_device_result_incorrect'),
           (0.0, 'patientt')]
          model.get_nearest_neighbors('syringe')
In [40]:
          [(0.9349400997161865, 'short'),
Out[40]:
           (0.9167957305908203, 'transferred'),
           (0.8911423087120056, 'vial'),
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           (0.8667225241661072, 'center'),
           (0.8654877543449402, 'spill,drug_injection'),
           (0.8617652654647827, 'spilt')]
          model.save_model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText_Vial_lyophilized
In [41]:
          model.save_model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText_Vial_lyophilized
```

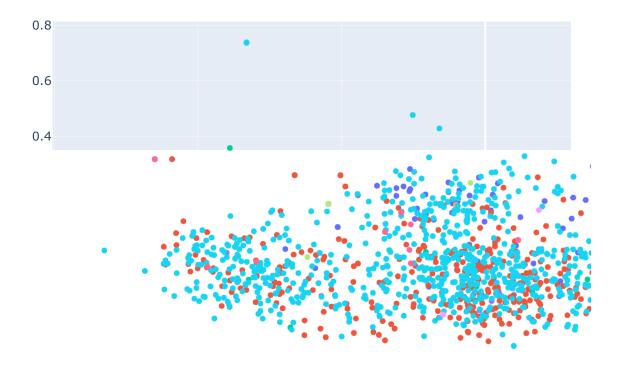
Software based device

```
In [42]:
    data = masterData.query(" name == 'Software based device' ").copy()
    valueCount = data["issue_type"].value_counts(normalize=True).to_frame().cumsum()*100
    data = data[data["issue_type"].isin(valueCount.index.tolist()[0:9])]
    data = data[["notes","issue_type","len"]]

    data['notes']=data['notes'].apply(lambda cw : clean_notes(cw))
    data = data.query(" len > 500 ")
```

hero.scatterplot(data, col="pca", color="issue_type", title="PCA issue_type")

PCA issue_type



```
In [44]: data["labeled_notes"] = data["issue_type"].apply(lambda x: '__label__' + x + " " ) + c
    x_train,x_test,y_train,y_test = train_test_split(data[["labeled_notes","issue_type"]],
    x_train.to_csv("C:/Users/gmodi/Downloads/x_train.csv",index=False,header=False)
    x_test.to_csv("C:/Users/gmodi/Downloads/x_test.csv",index=False,header=False)
    model = fasttext.train_supervised(input="C:/Users/gmodi/Downloads/x_train.csv", wordNgmodel.test("C:/Users/gmodi/Downloads/x_test.csv",k=3)

Out[44]: (916, 0.3296943231441048, 0.9890829694323144)

In [45]: # predict the data
    x_test["predicted"] = x_test["labeled_notes"].apply(lambda x: model.predict(x)[0][0]).
    print(classification_report(x_test["issue_type"], x_test["predicted"]))
    print(confusion_matrix(x_test["issue_type"], x_test["predicted"]))
    multiclass_roc_auc_score(x_test["issue_type"], x_test["predicted"])
```

```
precision
                                                           recall f1-score
                                                                                support
               Before Activation resolved
                                                   1.00
                                                              1.00
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      autoinjector_activation_difficulty
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                        customer feedback
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                            drug injection
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software based device connectivity issue
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              software based device other
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  software_based_device_result_incorrect
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                                                              0.72
   software based device technical issue
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                                                                        0.86
                                  accuracy
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                                                                        0.64
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                                 macro avg
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                              weighted avg
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```

C:\Users\gmodi\Anaconda3\envs\FastText\lib\site-packages\sklearn\metrics_classificat
ion.py:1318: UndefinedMetricWarning:

Precision and F-score are ill-defined and being set to 0.0 in labels with no predicte d samples. Use `zero_division` parameter to control this behavior.

C:\Users\gmodi\Anaconda3\envs\FastText\lib\site-packages\sklearn\metrics_classificat
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C:\Users\gmodi\Anaconda3\envs\FastText\lib\site-packages\sklearn\metrics_classificat
ion.py:1318: UndefinedMetricWarning:

Precision and F-score are ill-defined and being set to 0.0 in labels with no predicte d samples. Use `zero_division` parameter to control this behavior.

Out[45]: 0.8035630523220642

```
In [46]: x_test["prediction"] = x_test["labeled_notes"].apply(lambda x: model.predict(x,3)).ast
#x_test["prediction"] = x_test["prediction"].astype(str)
#x_test["prediction"] = x_test["prediction"].str.replace('__label__','')

x_test["predicted"] = ""
for i in range(len(x_test)):
    if x_test.issue_type.iloc[i] in x_test.prediction.iloc[i]: x_test.predicted.iloc[i]
    else: x_test.predicted.iloc[i] = 0

#x_test.to_csv("C:/Users/gmodi/Downloads/x_test_results.csv")
x_test["predicted"].value_counts(normalize=True)*100
```

```
Out[46]: 1 97.707424
0 2.292576
Name: predicted, dtype: float64

In [47]: model.save_model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText_SBD.bin")
model.save_model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText_SBD.ftz")
```

Tablet

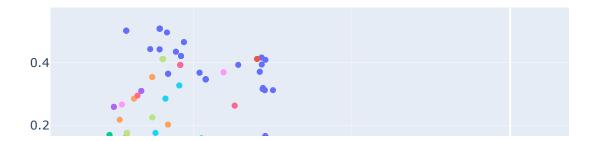
```
In [48]:
    data = masterData.query(" name == 'Tablet' ").copy()
    valueCount = data["issue_type"].value_counts(normalize=True).to_frame().cumsum()*100
    data = data[data["issue_type"].isin(valueCount.index.tolist()[0:9])]
    data = data[["notes","issue_type","len"]]

    data['notes']=data['notes'].apply(lambda cw : clean_notes(cw))
    data = data.query(" len > 500 ")

    data["notes"] = data["notes"].apply(lambda x: ' '.join(simple_preprocess(x, min_len=4, data["notes"] = data["notes"].apply(lambda x: remove_stopwords(''.join(x)))

In [49]:
    data["pca"] = (data["notes"].pipe(representation.tfidf, max_features=100).pipe(representation.scatterplot(data, col="pca", color="issue_type", title="PCA issue_type")
```

PCA issue type



```
data["labeled notes"] = data["issue type"].apply(lambda x: ' label ' + x + " " ) + c
In [50]:
          x_train,x_test,y_train,y_test = train_test_split(data[["labeled_notes","issue_type"]],
          x_train.to_csv("C:/Users/gmodi/Downloads/x_train.csv",index=False,header=False)
          x test.to csv("C:/Users/gmodi/Downloads/x test.csv",index=False,header=False)
          model = fasttext.train supervised(input="C:/Users/gmodi/Downloads/x train.csv", wordNg
          print(model.test("C:/Users/gmodi/Downloads/x_test.csv",k=3))
          # predict the data
          x test["predicted"] = x test["labeled notes"].apply(lambda x: model.predict(x)[0][0]).
          print(classification_report(x_test["issue_type"], x_test["predicted"]))
          print(confusion_matrix(x_test["issue_type"], x_test["predicted"]))
          multiclass_roc_auc_score(x_test["issue_type"], x_test["predicted"])
          (163, 0.3231083844580777, 0.9693251533742331)
                                        precision
                                                     recall f1-score
                                                                         support
                      To be determined
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              bottle damaged defective
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                 bottle induction seal
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                                                                  0.40
                 bottle label printing
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                       bottle quantity
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         carton cosmetic minor damage
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                                                                             101
              carton_damaged_defective
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                                                                               5
                     customer feedback
                                                                              36
                                             1.00
                                                       0.94
                                                                  0.97
                       drug appearance
                                                                               8
                                             0.60
                                                       0.38
                                                                  0.46
                                                                  0.92
                                                                             163
                              accuracy
                                             0.60
                                                       0.68
                                                                  0.61
                                                                             163
                             macro avg
                          weighted avg
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         C:\Users\gmodi\Anaconda3\envs\FastText\lib\site-packages\sklearn\metrics\ classificat
```

ion.py:1318: UndefinedMetricWarning:

Recall and F-score are ill-defined and being set to 0.0 in labels with no true sample s. Use `zero division` parameter to control this behavior.

C:\Users\gmodi\Anaconda3\envs\FastText\lib\site-packages\sklearn\metrics\ classificat ion.py:1318: UndefinedMetricWarning:

Recall and F-score are ill-defined and being set to 0.0 in labels with no true sample s. Use `zero division` parameter to control this behavior.

C:\Users\gmodi\Anaconda3\envs\FastText\lib\site-packages\sklearn\metrics_classificat ion.py:1318: UndefinedMetricWarning:

Recall and F-score are ill-defined and being set to 0.0 in labels with no true sample s. Use `zero division` parameter to control this behavior.

```
0.8758949467280674
Out[50]:
In [51]: x test["prediction"] = x test["labeled notes"].apply(lambda x: model.predict(x,3)).ast
         #x_test["prediction"] = x_test["prediction"].astype(str)
         #x_test["prediction"] = x_test["prediction"].str.replace('__label__','')
         x test["predicted"] = ""
          for i in range(len(x test)):
             if x test.issue type.iloc[i] in x test.prediction.iloc[i]: x test.predicted.iloc[i
             else: x_test.predicted.iloc[i] = 0
         #x test.to csv("C:/Users/qmodi/Downloads/x test results.csv")
         x test["predicted"].value counts(normalize=True)*100
              98.159509
Out[51]:
               1.840491
         Name: predicted, dtype: float64
         model.save model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText Tablet.bin")
In [52]:
         model.save model("C:/Users/gmodi/MyProjects/OtherDosageForms/FastText Tablet.ftz")
In [53]:
         PCM ISSUES, report codes = [],[]
In [54]: sampleRequest = """ <span style="font-size: 10pt;">Wellbean nurse </span><span styl</pre>
          clean notes(sampleRequest)
         'Caller stated they did not have further information to provide except injection site
Out[54]:
         was on patient s leg The activation button was pressed but the Sureclick pen did not
         work There were no click sound no needle penetration and no partial dose received fro
         m the complained unit No replacement is required '
 In [ ]: ##### Fast API Main Python File.
          from fastapi import FastAPI, HTTPException
          from pydantic import BaseModel
          import uvicorn
          import numpy as np
          import pandas as pd
          import re
          import nltk
          from bs4 import BeautifulSoup
          import fasttext
          from gensim.utils import simple preprocess
          from gensim.parsing.preprocessing import remove_stopwords
          def clean notes(text):
              soup = BeautifulSoup(text, 'html.parser')
             list1 = [item.get_text() for item in list(soup.children)]
             list2 = [i for i in list1 if len(i) == max([len(i) for i in list1])]
             list3 = [re.sub('[^a-zA-Z:]+', ' ', _) for _ in list2]
             return list3[0]
          # Declaring our FastAPI instance
          app = FastAPI()
         # Defining path operation for root endpoint
         @app.get("/")
         def main():
```

```
return {
        "message": "Welcome to Amgen AI!"
class request_body(BaseModel):
    AutomationId: str
    DosageForm: str
    Product: str
    ProductID: str
    MasterCase: str
    PCM Subcase: str
    OccurCountry: str
    PPQ: str
    Notes: str
@app.post("/AMD")
def AMD(data: request_body):
    amd model = fasttext.load model("FastText AMD.ftz")
    issuePredicted = amd_model.predict(clean_notes(data.Notes),k=3)
    PCM ISSUES, report codes = [],[]
    for j in range(3):
        report_codes.append({'reported_code': issuePredicted[0][j].replace('__label__
    PCM_ISSUES.append({'verbatim': list3[0], 'report_codes': report_codes})
    return {
        "AutomationId": data.AutomationId,
        "DosageForm": data.DosageForm,
        "Product": data.Product,
        "ProductID": data.ProductID,
        "MasterCase": data.MasterCase,
        "PCM_Subcase": data.PCM_Subcase,
        "OccurCountry": data.OccurCountry,
        "PPQ": data.PPQ,
        "PCM_ISSUES": PCM_ISSUES
     }
```