## CMPE 537 - 2020 Fall

# Image Classification with Bag-of-Features

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## 1 Methodology

### 1.1 Keypoint Extraction

We use *Bag-of-Features* approach to classify images. In order to extract *features*, we use SIFT (Scale-Invariant Feature Transform). SIFT algorithm outputs keypoints which has a 128 dimensional descriptor.

### 1.2 Keypoints to Features

In order to convert keypoints to features, we use clustering. Then, we use found K cluster centers to convert keypoints to features (Quantization). In this experiment, we choose K-Means. Nevertheless, other cluster algorithms can also be replaced instead.

### 1.3 Bag-of-Features

Now, we use extracted features to create a histogram of an image. (Bag-of-Features) For that purpose, we count the occurrences of each feature in an image. As the result, we get K dimensional vector where each value represents how many times that feature occurred in that image.

### 1.4 Classification

As we can represent each image with a  $\boldsymbol{K}$  dimensional vector, we use these data points to classify the data set. We can choose any classification algorithm, but for this experiment, we used Support Vector Classification with Chi-Squared kernel.

### 2 Experiment Setup

### 2.1 Dataset

For this experiment, we use Caltech20 datasets, which contains 20 class with additional reject class. (background\_class) Classes do not contain equal amount of samples.

### 2.2 Parameters

There are several parameters that we will try to optimize.

First one is C value. C value determines the weakness of the regularization, which is squared L2.

Then, we will check if the model performs better or worse when we implement Chi-Squared kernel.

Lastly, we will observe the effect of  $\boldsymbol{K}$ . We will test the model with 50,100 and 500 clusters.

### 2.3 Procedure & Evaluation

For each combination, we will use K-Fold cross verification with 5 partitions on training set. Then, using found  $y_{pred}$  and  $y_{true}$ , we calculate metrics.

The metrics include, F1 score for each class, F1 score mean, and confusion matrix. As we find the best combination of parameters, we will use that model to test set to get the final evaluation.

### 3 Results

We ran total of 24 tests. Parameters for each test is indicated in table below with resulting F1 macro score from 5-Fold cross validation. Chi<sup>2</sup> kernel is applied with  $\gamma$  value 0.5.

#	K	Kernel	$\mathbf{C}$	F1
1	50	Linear	1	0.02
<b>2</b>	50	Linear	10	0.03
3	50	Linear	20	0.04
4	50	Linear	30	0.04
<b>5</b>	50	$\mathrm{Chi}^2$	1	0.11
6	50	$\mathrm{Chi}^2$	10	0.33
7	50	$\mathrm{Chi}^2$	20	0.43
8	50	$\mathrm{Chi}^2$	30	0.47
9	100	Linear	1	0.02
10	100	Linear	10	0.03
11	100	Linear	20	0.04
12	100	Linear	30	0.04
<b>13</b>	100	$\mathrm{Chi}^2$	1	0.12
<b>14</b>	100	$\mathrm{Chi}^2$	10	0.46
15	100	$\mathrm{Chi}^2$	20	0.53
<b>16</b>	100	$\mathrm{Chi}^2$	30	0.56
<b>17</b>	500	Linear	1	0.02
18	500	Linear	10	0.03
<b>19</b>	500	Linear	20	0.03
<b>20</b>	500	Linear	30	0.04
21	500	$\mathrm{Chi}^2$	1	0.15
<b>22</b>	500	$\mathrm{Chi}^2$	10	0.61
<b>23</b>	500	$\mathrm{Chi}^2$	20	0.63
24	500	$\mathrm{Chi}^2$	30	0.63

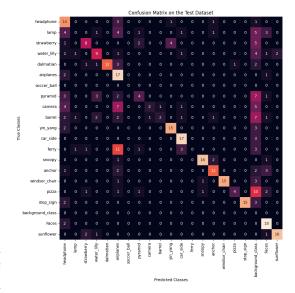
As it can be observed, when C and K values increase, results are improving. Also, Chi<sup>2</sup> kernel has significant advantage over linear kernel. Full results are shared in Appendix, with per-class F1 scores, and confusion matrices.

### 3.1 Best Model on the Test Set

Best parameters are set to be: K=500,  $Chi^2$ , C=30. After a model with this parameters is trained again, we get 0.64 F1 macro score on cross validation. However, at the test set, model results **0.49** F1 macro score with other scores below.

Class	F1 Score
headphone	0.48
lamp	0.00
strawberry	0.48
${\bf water\_lilly}$	0.49
dalmatian	0.75
airplanes	0.45
$\mathbf{soccer\_ball}$	0.00
pyramid	0.28
camera	0.17

barrel	0.17
yinyang	0.75
$\operatorname{car\_side}$	0.77
ferry	0.00
snoopy	0.84
anchor	0.65
$windsor\_chair$	0.86
pizza	0.32
$\operatorname{stop\_sign}$	0.86
$background\_class$	0.00
Faces	0.69
sunflower	0.84



### 3.2 Mislabeled Samples

As F1 scores are very low, this method can not be applied for critical applications such as self-driving cars. In this sections, we also observe stop signs that are mislabeled.



True: Ferry Predicted: Airplane



True: Ferry Predicted: Background Class



True: Water Lilly Predicted: Background Class



True: Strawberry Predicted: Yin Yang



True: Camera Predicted: Headphone



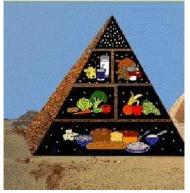
True: Camera Predicted: Headphone



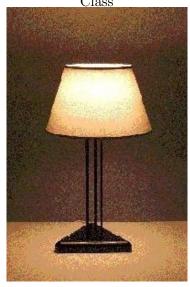
True: Lamp Predicted: Background Class



True: Pyramid Predicted: Airplanes



True: Pyramid Predicted: Background Class



True: Lamp Predicted: Headphone



True: Car Side **Predicted:** Background Class



True: Water Lilly **Predicted:** Background



True: Stop Sign Predicted: Background Class



True: Camera Predicted: Airplane



True: Ferry Predicted: Background Class



True: Barrel Predicted: Camera



True: Barrel Predicted: Airplane



True: Ferry Predicted: Car Side



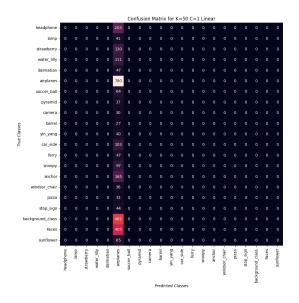
True: Ferry Predicted: Airplane



True: Stop Sign Predicted: Background Class

# 4 Appendix

## 4.1 K=50 C=1 Linear

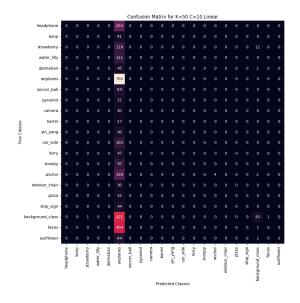


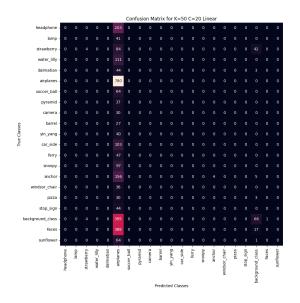
**F1 Macro Score** : 0.021

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.00
${f water\_lilly}$	0.00
dalmatian	0.00
airplanes	0.42
$\mathbf{soccer\_ball}$	0.00
$\mathbf{pyramid}$	0.00
camera	0.00
barrel	0.00
${f yin\_yang}$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.00
${f windsor\_chair}$	0.00
pizza	0.00
$\operatorname{stop\_sign}$	0.00
$background\_class$	0.02
Faces	0.00
sunflower	0.00

## 4.2 K=50 C=10 Linear

## 4.3 K=50 C=20 Linear





F1 Macro Score: 0.030

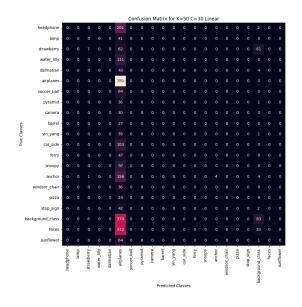
F1 Macro Score: 0.036

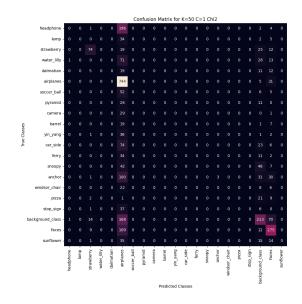
Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.00
$\mathbf{water\_lilly}$	0.00
dalmatian	0.00
airplanes	0.42
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$\mathbf{yin}_{-}\mathbf{yang}$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.05
$\mathbf{windsor}_{-}\mathbf{chair}$	0.00
pizza	0.00
$\mathbf{stop\_sign}$	0.00
$background\_class$	0.16
Faces	0.00
$\operatorname{sunflower}$	0.00

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.06
water_lilly	0.00
dalmatian	0.00
airplanes	0.43
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$yin_yang$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.05
$windsor\_chair$	0.00
pizza	0.00
$\operatorname{stop\_sign}$	0.00
$background\_class$	0.22
Faces	0.00
sunflower	0.00

## 4.4 K=50 C=30 Linear

# 4.5 K=50 C=1 Chi-Squared





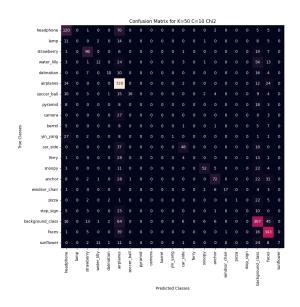
**F1 Macro Score** : 0.040

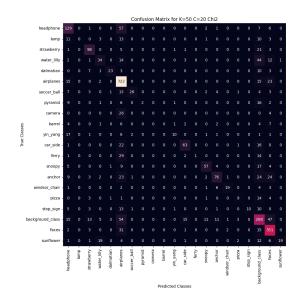
F1 Macro Score: 0.208

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.10
$\mathbf{water\_lilly}$	0.00
dalmatian	0.00
airplanes	0.44
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$\mathbf{yin}_{-}\mathbf{yang}$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.05
$\mathbf{windsor}_{-}\mathbf{chair}$	0.00
pizza	0.00
$\mathbf{stop\_sign}$	0.00
$background\_class$	0.25
Faces	0.00
$\operatorname{sunflower}$	0.00

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.62
$water\_lilly$	0.00
dalmatian	0.00
airplanes	0.56
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$yin_yang$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.04
$windsor\_chair$	0.00
pizza	0.00
${f stop\_sign}$	0.00
$background\_class$	0.45
Faces	0.60
sunflower	0.00

# K=50 C=10 Chi-Squared 4.7 K=50 C=20 Chi-Squared





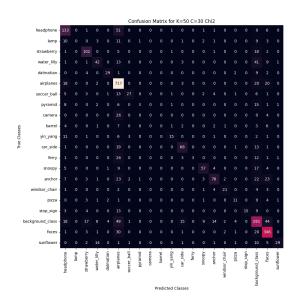
F1 Macro Score: 0.329

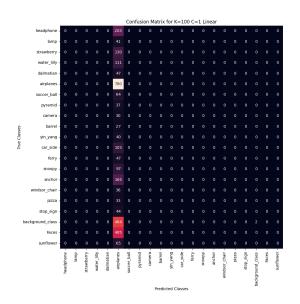
**F1 Macro Score** : 0.438

Class	F1 Score
headphone	0.55
lamp	0.00
strawberry	0.72
${f water\_lilly}$	0.17
dalmatian	0.32
airplanes	0.75
$\mathbf{soccer\_ball}$	0.40
pyramid	0.00
camera	0.00
barrel	0.00
$\mathbf{yin}_{-}\mathbf{yang}$	0.05
$\operatorname{car\_side}$	0.57
ferry	0.00
snoopy	0.65
anchor	0.55
$windsor\_chair$	0.63
pizza	0.06
$\mathbf{stop\_sign}$	0.00
$background\_class$	0.57
Faces	0.74
sunflower	0.19

Class	F1 Score
headphone	0.59
lamp	0.00
strawberry	0.73
$water\_lilly$	0.38
dalmatian	0.61
airplanes	0.79
$soccer\_ball$	0.55
pyramid	0.10
camera	0.00
barrel	0.00
$yin_yang$	0.37
$\operatorname{car\_side}$	0.65
ferry	0.04
snoopy	0.65
anchor	0.56
$windsor\_chair$	0.64
pizza	0.37
$\operatorname{stop\_sign}$	0.37
$background\_class$	0.57
Faces	0.77
sunflower	0.45

# K=50 C=30 Chi-Squared 4.9 K=100 C=1 Linear





F1 Macro Score: 0.477

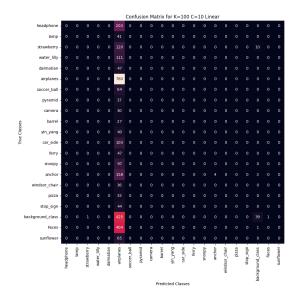
**F1 Macro Score** : 0.0202

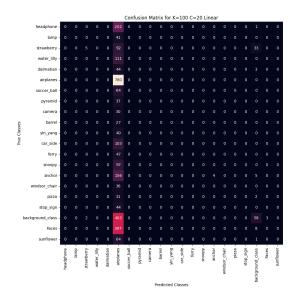
Class	F1 Score
headphone	0.62
lamp	0.00
$\operatorname{strawberry}$	0.74
${f water\_lilly}$	0.45
dalmatian	0.69
airplanes	0.79
$\mathbf{soccer\_ball}$	0.55
pyramid	0.14
camera	0.00
barrel	0.00
${f yin\_yang}$	0.50
$\operatorname{car}$ _side	0.68
ferry	0.12
snoopy	0.66
anchor	0.57
${f windsor\_chair}$	0.66
pizza	0.41
$\operatorname{stop\_sign}$	0.51
$background\_class$	0.56
Faces	0.78
$\operatorname{sunflower}$	0.59

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.00
$water\_lilly$	0.00
dalmatian	0.00
airplanes	0.42
$soccer\_ball$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
yin_yang	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.00
$windsor\_chair$	0.00
pizza	0.00
$stop\_sign$	0.00
$background\_class$	0.01
Faces	0.00
sunflower	0.00

# 4.10 K=100 C=10 Linear

## 4.11 K=100 C=20 Linear





**F1 Macro Score** : 0.0295

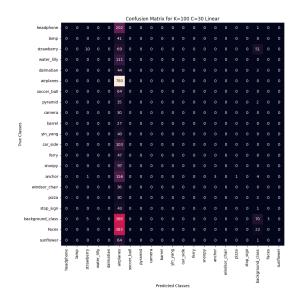
**F1 Macro Score** : 0.0358

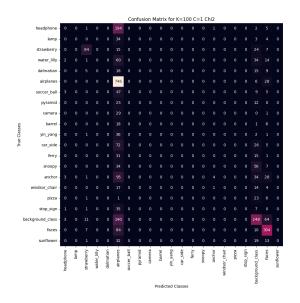
Class	F1 Score
headphone	0.00
lamp	0.00
$\operatorname{strawberry}$	0.00
${f water\_lilly}$	0.00
dalmatian	0.00
airplanes	0.42
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
${f yin\_yang}$	0.00
$\operatorname{car}$ _side	0.00
ferry	0.00
snoopy	0.00
anchor	0.00
${f windsor\_chair}$	0.00
pizza	0.00
$\operatorname{stop\_sign}$	0.00
$background\_class$	0.15
Faces	0.00
$\operatorname{sunflower}$	0.00

Class	F1 Score
	0.00
headphone	
lamp	0.00
strawberry	0.07
${f water\_lilly}$	0.00
dalmatian	0.00
airplanes	0.43
$soccer\_ball$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$yin_yang$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.05
$windsor\_chair$	0.00
pizza	0.00
$stop\_sign$	0.00
$background\_class$	0.20
Faces	0.00
sunflower	0.00

## 4.12 K=100 C=30 Linear

# 4.13 K=100 C=1 Chi-Squared





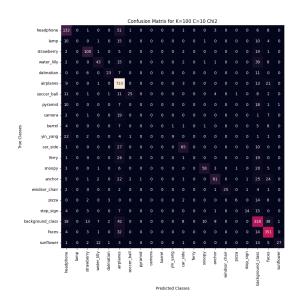
**F1 Macro Score** : 0.0397

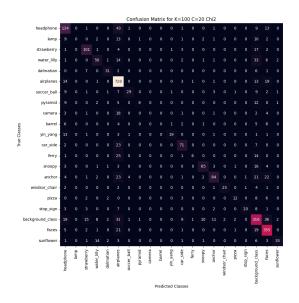
F1 Macro Score: 0.117

Class	F1 Score
headphone	0.00
lamp	0.00
$\operatorname{strawberry}$	0.14
${f water\_lilly}$	0.00
dalmatian	0.00
airplanes	0.44
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$\mathbf{yin}_{-}\mathbf{yang}$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.04
$\mathbf{windsor}_{-}\mathbf{chair}$	0.00
pizza	0.00
$\mathbf{stop\_sign}$	0.00
$background\_class$	0.22
Faces	0.00
sunflower	0.00

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.69
$water\_lilly$	0.00
dalmatian	0.00
airplanes	0.59
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$yin_yang$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.05
$windsor\_chair$	0.00
pizza	0.00
$\mathbf{stop\_sign}$	0.00
$background\_class$	0.48
Faces	0.66
sunflower	0.00

# 4.14 K=100 C=10 Chi-Squared 4.15 K=100 C=20 Chi-Squared





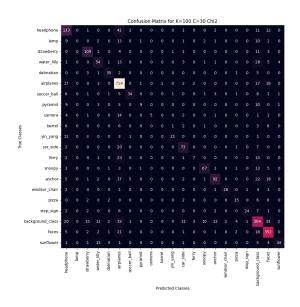
**F1 Macro Score** : 0.459

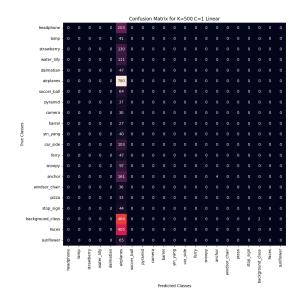
F1 Macro Score: 0.531

Class	F1 Score
headphone	0.59
lamp	0.00
strawberry	0.75
${f water\_lilly}$	0.48
dalmatian	0.60
airplanes	0.81
$\mathbf{soccer\_ball}$	0.54
pyramid	0.00
camera	0.00
barrel	0.00
$\mathbf{yin}_{-}\mathbf{yang}$	0.36
$\operatorname{car\_side}$	0.68
ferry	0.00
snoopy	0.69
anchor	0.60
$windsor\_chair$	0.82
pizza	0.29
$\operatorname{stop\_sign}$	0.47
$background\_class$	0.61
Faces	0.78
sunflower	0.57

Class	F1 Score
Class	TT Score
headphone	0.61
lamp	0.00
strawberry	0.76
$water\_lilly$	0.52
dalmatian	0.71
airplanes	0.83
$soccer\_ball$	0.58
pyramid	0.34
camera	0.00
barrel	0.00
yin_yang	0.61
$\operatorname{car\_side}$	0.70
ferry	0.22
snoopy	0.73
anchor	0.61
$windsor\_chair$	0.79
pizza	0.48
$stop\_sign$	0.61
$background\_class$	0.63
Faces	0.80
sunflower	0.63

# 4.16 K=100 C=30 Chi-Squared 4.17 K=500 C=1 Linear





F1 Macro Score: 0.565

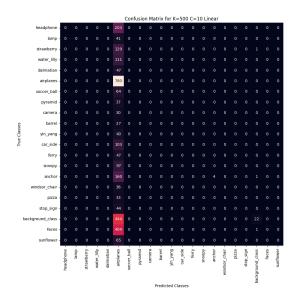
**F1 Macro Score** : 0.0225

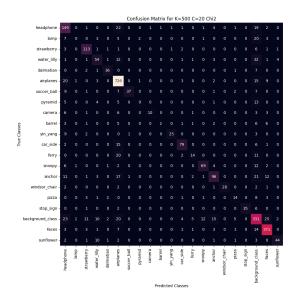
Class	F1 Score
headphone	0.60
lamp	0.00
strawberry	0.80
${f water\_lilly}$	0.52
dalmatian	0.75
airplanes	0.83
$\mathbf{soccer\_ball}$	0.64
pyramid	0.37
camera	0.29
barrel	0.00
$\mathbf{yin}_{-}\mathbf{yang}$	0.66
$\operatorname{car}$ _side	0.71
ferry	0.24
snoopy	0.74
anchor	0.65
${f windsor\_chair}$	0.81
pizza	0.52
$\operatorname{stop\_sign}$	0.69
$background\_class$	0.63
Faces	0.80
$\operatorname{sunflower}$	0.64

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.00
$water\_lilly$	0.00
dalmatian	0.00
airplanes	0.42
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$yin_yang$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.05
$windsor\_chair$	0.00
pizza	0.00
${f stop\_sign}$	0.00
$background\_class$	0.01
Faces	0.00
sunflower	0.00

## 4.18 K=500 C=10 Linear

## $4.19 ext{ K=}500 ext{ C=}20 ext{ Linear}$



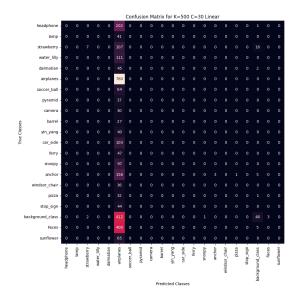


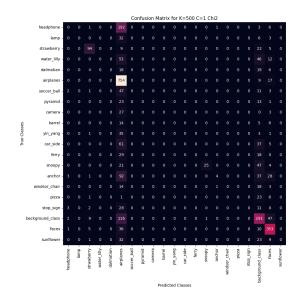
F1 Macro Score: 0.0265

F1 Macro Score: 0.0304

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.00
$\mathbf{water\_lilly}$	0.00
dalmatian	0.00
airplanes	0.42
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$\mathbf{yin}_{-}\mathbf{yang}$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.05
${\bf windsor\_chair}$	0.00
pizza	0.00
$\mathbf{stop\_sign}$	0.00
$background\_class$	0.09
Faces	0.00
$\operatorname{sunflower}$	0.00

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.00
$water\_lilly$	0.00
dalmatian	0.00
airplanes	0.42
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$yin_yang$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.05
$windsor\_chair$	0.00
pizza	0.00
$\operatorname{stop\_sign}$	0.00
$background\_class$	0.17
Faces	0.00
sunflower	0.00





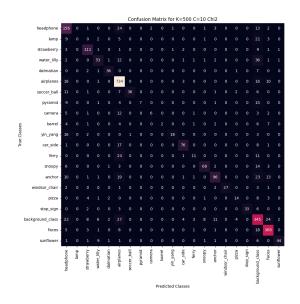
**F1 Macro Score** : 0.0352

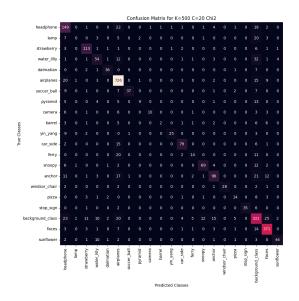
F1 Macro Score: 0.148

Class	F1 Score
headphone	0.00
lamp	0.00
$\operatorname{strawberry}$	0.10
${f water\_lilly}$	0.00
dalmatian	0.00
${f airplanes}$	0.43
$\operatorname{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$\mathbf{yin}_{-}\mathbf{yang}$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.00
anchor	0.04
$windsor\_chair$	0.00
pizza	0.00
$\operatorname{stop\_sign}$	0.00
$background\_class$	0.18
Faces	0.00
sunflower	0.00

Class	F1 Score
headphone	0.00
lamp	0.00
strawberry	0.75
water_lilly	0.00
dalmatian	0.00
airplanes	0.62
$\mathbf{soccer\_ball}$	0.00
pyramid	0.00
camera	0.00
barrel	0.00
$yin_yang$	0.00
$\operatorname{car\_side}$	0.00
ferry	0.00
snoopy	0.41
anchor	0.05
$windsor\_chair$	0.00
pizza	0.00
$\operatorname{stop\_sign}$	0.00
$background\_class$	0.52
Faces	0.76
sunflower	0.00

# 4.22 K=500 C=10 Chi-Squared 4.23 K=500 C=20 Chi-Squared





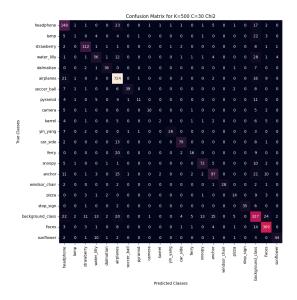
F1 Macro Score: 0.614

F1 Macro Score: 0.631

Class	F1 Score
headphone	0.65
lamp	0.00
strawberry	0.83
${f water\_lilly}$	0.56
dalmatian	0.80
airplanes	0.87
$\mathbf{soccer\_ball}$	0.71
pyramid	0.32
camera	0.32
barrel	0.14
$\mathbf{yin}_{-}\mathbf{yang}$	0.60
$\operatorname{car\_side}$	0.80
ferry	0.33
snoopy	0.76
anchor	0.67
$windsor\_chair$	0.86
pizza	0.52
$\mathbf{stop\_sign}$	0.86
$background\_class$	0.66
Faces	0.87
sunflower	0.78

Class	F1 Score
Class	TT SCOTC
headphone	0.64
lamp	0.00
strawberry	0.83
$water\_lilly$	0.53
dalmatian	0.79
airplanes	0.88
$soccer\_ball$	0.72
pyramid	0.37
camera	0.49
barrel	0.13
yin_yang	0.75
$\operatorname{car\_side}$	0.80
ferry	0.39
snoopy	0.76
anchor	0.64
$windsor\_chair$	0.88
pizza	0.48
$stop\_sign$	0.89
$background\_class$	0.65
Faces	0.88
sunflower	0.76

# 4.24~ K=500 C=30 Chi-Squared



F1 Macro Score: 0.640

Class	F1 Score
headphone	0.65
lamp	0.04
$\operatorname{strawberry}$	0.83
$\mathbf{water\_lilly}$	0.54
dalmatian	0.79
airplanes	0.88
$\mathbf{soccer\_ball}$	0.74
pyramid	0.43
camera	0.48
barrel	0.12
$\mathbf{yin}_{-}\mathbf{yang}$	0.76
$\operatorname{car}$ -side	0.81
ferry	0.43
snoopy	0.76
anchor	0.64
$\mathbf{windsor}_{\mathtt{c}}\mathbf{chair}$	0.88
pizza	0.48
${f stop\_sign}$	0.89
$background\_class$	0.65
Faces	0.88
$\operatorname{sunflower}$	0.76