<pre>In [52]: Out[52]: In [53]: In [54]:</pre>	<pre>import matplotlib as plt import seaborn as sns import pandas as pd import numpy as np import json as js</pre>
In [54]:	'/Users/mingyuanma/Desktop/HAI/analysis'
	<pre>before = before.rename(columns = {"label":"label_dim1"})  before_dim2 = pd.read_csv("//data/phasel/before_dim2.csv") before_dim2 = before_dim2.rename(columns = {"Unnamed: 2":"label"}).loc[:,["ResponseId", "label"]] before = before_dim2.merge(before, how="right", left_on="ResponseId", right_on="ResponseId") before = before[before["round"] &gt;= 3]</pre>
Out[54]:	Responseld label round tick orderld taskId taskTicks workerId workerTicks label_dim1           135 R_0uOsLe6BeLnUee5 3 3 3 1 2 1 1 1 1 2 1 0 0 0 0 0 0 0 0 0 0 0 0
	138       R_OuOsLe6BeLnUee5       3       3       2       0       0       0       0       0       3         139       R_OuOsLe6BeLnUee5       3       3       2       0       0       0       0       0       3                     100561       R_zTlbYwl8ijcujhn       7       6       33       0       0       0       0       0       1
	100562       R_zTlbYwl8ijcujhn       7       6       33       0       0       0       0       0       1         100563       R_zTlbYwl8ijcujhn       7       6       34       4       3       2       1       0       1         100564       R_zTlbYwl8ijcujhn       7       6       34       0       0       0       0       0       1         100565       R_zTlbYwl8ijcujhn       7       6       34       0       0       0       0       1         76941 rows × 10 columns       10       10       0       0       0       0       0       0       0       0
In [55]: In [56]:	<pre>after = pd.read_csv("//data/phase2/combined2.csv") after = after.rename(columns = {"label":"label_dim1"})  after_dim2 = pd.read_csv("//data/phase2/after_dim2.csv")</pre>
Out[56]:	<pre>after_dim2 = after_dim2.rename(columns = {"Unnamed: 2":"label"}).loc[:,["ResponseId", "label"]] after = after_dim2.merge(after, how="right", left_on="ResponseId", right_on="ResponseId") after = after[after["round"] &gt;= 3] after  ResponseId label round tick orderld taskId taskTicks workerId workerTicks label_dim1  117 R_0HBxBV8U696D9QJ 4 3 1 1 1 2 0 0 2</pre>
	118       R_OHBxBV8U696D9QJ       4       3       1       2       1       0       2         119       R_OHBxBV8U696D9QJ       4       3       1       0       0       0       0       2         120       R_OHBxBV8U696D9QJ       4       3       2       3       1       2       1       0       2         121       R_OHBxBV8U696D9QJ       4       3       2       0       0       0       0       2
	.
	141761       R_zZQIbE0LFD13yRX       5       6       38       0       0       0       0       0       3         108330 rows × 10 columns       Chi-square testing
In [57]:	
	<pre>stat, p, dof, expected = chi2_contingency(data)  else:     c1, c2 = Counter(array1), Counter(array2)     before_dis, after_dis = [], []     for i in set(c1).union(set(c2)):         before_dis.append(c1[i])         after_dis.append(c2[i])</pre>
	<pre>print(before_dis) print(after_dis) data = [before_dis, after_dis] stat, p, dof, expected = chi2_contingency(data)  alpha = 0.05 print("p value is " + str(p))</pre>
	if p <= alpha:     print('difference between the two distributions (reject H0)') else:     print('no difference between the two distributions (H0 holds true)')  Distribution of label before and after
In [58]: In [59]:	d1 = before.groupby( ResponseId , as_index=False).agg(lambda x.x.iloc[0]) d2 = after.groupby("ResponseId", as_index=False).agg(lambda x:x.iloc[0])
	0.35 - 0.30 - 0.25 - 19 0.20 - 20 0.15 -
	8 0.15 - 0.10 - 0.05 - 0.00 - 1 2 3 4 5 6 7 8 9
In [60]:	sns.histplot(data=d2, x="label", stat="probability", color="skyblue");
	0.30 - 10 - 10 - 10 - 10 - 10 - 10 - 10 - 10 -
	Chi-Square Testing
In [61]:	chisquare(d1["label"], d2["label"])  [16, 0, 62, 26, 24, 14, 11, 4, 13] [32, 2, 95, 17, 58, 6, 5, 1, 29] p value is 0.0001373003673541072 difference between the two distributions (reject H0)
In [62]:	<pre>ids = [] rounds = [] server = []</pre>
	<pre>label = [] for player in set(np.array(df["ResponseId"])):     for i in np.arange(1,7):         tem = df[(df["ResponseId"] == player) &amp; (df["round"] == i)]         if len(tem) != 0: # no response there             ids.append(player)             rounds.append(i)         l = df[(df["ResponseId"] == player)]["label"].iloc[0]</pre>
	<pre>label.append(1) if i &lt;= 2:     num = sum((tem["workerId"] == 2) &amp; (tem["taskId"] == 1)) else:     num = sum((tem["workerId"] == 1) &amp; (tem["taskId"] == 1))     server.append(num) # else: # print(player, i)</pre>
	<pre>d = {     "ResponseId": ids,     "round": rounds,     "numServerCook": server,     "label": label } return pd.DataFrame(d)</pre>
In [63]: In [64]: In [65]:	after_compliance = compliance(after)
In [65]: Out[65]:	Responseld round numServerCook label  Responseld vound numServerCook label
	3       R_8nOXycxlDgZm6kx       6       2       5         4       R_6JZr4lgBJ3Xo9gt       3       1       5                 675       R_BYxHoWpYxArK6dP       6       3       4
	676       R_1KiNAbM6I7sMKtU       3       2       3         677       R_1KiNAbM6I7sMKtU       4       2       3         678       R_1KiNAbM6I7sMKtU       5       2       3         679       R_1KiNAbM6I7sMKtU       6       2       3         680 rows × 4 columns
In [66]: Out[66]:	after_compliance  Responseld round numServerCook label  R_1MX11huDrBtkGoB 3 2 5
	1       R_1MX11huDrBtkGoB       4       2       5         2       R_1MX11huDrBtkGoB       5       2       5         3       R_1MX11huDrBtkGoB       6       2       5         4       R_1jYXGl34FVbha4S       3       2       3
	975       R_ykkhkGYu1KpIM8h       6       4       3         976       R_3gMLtEVUqUbNFXW       3       10       4         977       R_3gMLtEVUqUbNFXW       4       3       4         978       R_3gMLtEVUqUbNFXW       5       1       4
In [ ]:	979 R_3gMLtEVUqUbNFXW 6 3 4 980 rows × 4 columns
In [67]: Out[67]:	round_before
Ouc[67]:	round       3     2.552941       4     2.670588       5     2.617647
In [68]: Out[68]:	round_after
	round       3     2.408163       4     2.232653       5     2.163265       6     2.044898
In [69]:	a - mp.array(round_before[ numbervercook ])
In [71]:	chi-square testing  chisquare(a, b, count=False)  p value is 0.9994561534639331  no difference between the two distributions (HO holds true)
In [ ]: In [72]:	group by tip
Out[72]:	tip_before    label   numServerCook
	<ul> <li>2 4 2.403846</li> <li>3 5 2.479167</li> <li>4 6 2.571429</li> <li>5 7 2.454545</li> <li>6 8 2.687500</li> </ul>
In [73]: Out[73]:	tip_after  tip_after
	<ul> <li>1 1.914062</li> <li>1 2 2.375000</li> <li>2 3 2.215789</li> <li>3 4 2.382353</li> <li>4 5 2.219828</li> </ul>
	5       6       2.166667         6       7       1.650000         7       8       2.000000         8       9       2.517241
In [74]:	sns.barplot(x="label", y="numServerCook", data=tip_before, alpha=0.8);
	No       2.0 -       1.5 -
In [75]:	0.5 - 0.0 1 3 4 5 6 7 8 9
In [75]:	sns.barplot(x="label", y="numServerCook", data=tip_after, alpha=0.8);  25  20  21  25  21  25  26  27  28  29  20  20  20  20  20  20  20  20  20
In [75]:	05 - 00 1 3 4 5 6 7 8 9  sns.barplot(x="label", y="numServerCook", data=tip_after, alpha=0.8);  25 - 20 - 00 - 00 - 00 - 00 - 00 - 00 -
<pre>In [75]: Out[76]:</pre>	sns.barplot(x="label", y="numServerCook", data=tip_efter, alpha=0.8);  25  20  20  21  25  26  27  28  28  29  20  20  20  20  20  20  20  20  20
In [76]:	### chi-square testing    ans.barplot(x="label", y="numServerCook", data=tip_after, alpha=0.8);    chi-square testing
In [76]: Out[76]:	chi-square testing  property (1.403816, 2.4392666, 2.4036615, 2.47916667, 2.57142857, 2.4545368, 2.6375, 2.2853556, 2.4036615, 2.47916667, 2.57142857, 2.4645368, 2.6375, 2.2863667)  property (1.402811, 2.2928981), 2.00737376, 1.96484444, 0, 2.15740741, 2.4166867, 2.05355596, 0)  property (1.402811, 2.2928981), 2.00737376, 1.96484444, 0, 2.15740741, 2.4166867, 2.05355596, 0)  property (1.402811, 2.2928981), 2.00737376, 1.96484444, 0, 2.15740741, 2.4166867, 2.05355596, 0)  property (1.402811, 2.2928981), 2.00737376, 1.96484444, 0, 2.15740741, 2.4166867, 2.05355596, 0)  property (1.402811, 2.2928981), 2.00737376, 1.96484444, 0, 2.15740741, 2.4166867, 2.05355596, 0)  Testing Aversion Distribution
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