

# DATA 221 Homework 2

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## Problem 1

### 1(a)

Calculate

$$P(\text{WB score} = \text{Improved}, \text{PCC symptoms} = \text{Less} \mid \text{Vaccinated})$$

```
import pandas as pd

train = pd.read_csv("data/PCC_study_train.csv")

vacc = train[train["vax_status"] == "Vaccinated"]

p_vacc = (
    (vacc["WBscore"] == "Improved") &
    (vacc["PCCsymp"] == "Less")
).mean()

print(p_vacc)
```

0.75

### 1(b)

Calculate

$$P(\text{WB score} = \text{Improved}, \text{PCC symptoms} = \text{Less} \mid \text{Unvaccinated})$$

```

unvacc = train[train["vax_status"] == "Unvaccinated"]

p_unvacc = (
    (unvacc["WBscore"] == "Improved") &
    (unvacc["PCCsymp"] == "Less")
).mean()

print(p_unvacc)

```

0.07692307692307693

**1(c)**

```

import pandas as pd

def make_table(df, label):
    wb = (
        df["WBscore"]
        .value_counts(normalize=True)
        .rename("Probability")
        .reset_index()
        .rename(columns={"index": "WBscore"})
    )

    pcc = (
        df["PCCsymp"]
        .value_counts(normalize=True)
        .rename("Probability")
        .reset_index()
        .rename(columns={"index": "PCCsymp"})
    )

    print(f"\n{n[label]} - Well-being score")
    display(wb)

    print(f"\n{n[label]} - PCC symptoms")
    display(pcc)

```

```
make_table(vacc, "Vaccinated")
make_table(unvacc, "Unvaccinated")
```

Vaccinated - Well-being score

	WBscore	Probability
0	Improved	0.777778
1	Unchanged	0.138889
2	Worsened	0.083333

Vaccinated - PCC symptoms

	PCCsymp	Probability
0	Less	0.861111
1	Same	0.083333
2	More	0.055556

Unvaccinated - Well-being score

	WBscore	Probability
0	Unchanged	0.512821
1	Worsened	0.307692
2	Improved	0.179487

Unvaccinated - PCC symptoms

	PCCsymp	Probability
0	Same	0.512821
1	More	0.282051

	PCCsymp	Probability
2	Less	0.205128

### 1(d)

Assuming conditional independence, write a function that computes

$$P(\text{outcome} \mid \text{class})$$

```
def naive_bayes_likelihood(
    wb_score,
    pcc_symp,
    wb_probs,
    pcc_probs,
    class_prob
):
    """
    wb_score: The observed WBScore (e.g., "Improved")
    pcc_symp: The observed PCCsymp (e.g., "Less")
    wb_probs: Dictionary of conditional probs for WBScore given the class
    pcc_probs: Dictionary of conditional probs for PCCsymp given the class
    class_prob: The prior probability of the class (e.g., P(Vaccinated))
    """
    return (
        wb_probs[wb_score] *
        pcc_probs[pcc_symp] *
        class_prob
    )
```

### 1(e)

```
p_vaccinated = (train["vax_status"] == "Vaccinated").mean()
p_unvaccinated = (train["vax_status"] == "Unvaccinated").mean()

vacc_wb_probs = vacc["WBScore"].value_counts(normalize=True)
vacc_pcc_probs = vacc["PCCsymp"].value_counts(normalize=True)

unvacc_wb_probs = unvacc["WBScore"].value_counts(normalize=True)
unvacc_pcc_probs = unvacc["PCCsymp"].value_counts(normalize=True)
```

```

# likelihoods for WB = Worsened, PCC = Same
lik_vacc = naive_bayes_likelihood(
    "Worsened",
    "Same",
    vacc_wb_probs,
    vacc_pcc_probs,
    p_vaccinated
)

lik_unvacc = naive_bayes_likelihood(
    "Worsened",
    "Same",
    unvacc_wb_probs,
    unvacc_pcc_probs,
    p_unvaccinated
)

print("Vaccinated likelihood:", lik_vacc)
print("Unvaccinated likelihood:", lik_unvacc)

```

```

Vaccinated likelihood: 0.003333333333333333
Unvaccinated likelihood: 0.08205128205128205

```

**1(f)**

```

test = pd.read_csv("data/PCC_study_test.csv")

def predict_vax_status(row):
    lv = naive_bayes_likelihood(
        row["WBscore"],
        row["PCCsymp"],
        vacc_wb_probs,
        vacc_pcc_probs,
        p_vaccinated
    )
    lu = naive_bayes_likelihood(
        row["WBscore"],
        row["PCCsymp"],
        unvacc_wb_probs,

```

```

        unvacc_pcc_probs,
        p_unvaccinated
    )
    return "Vaccinated" if lv > lu else "Unvaccinated"

test["predicted_vax_status"] = test.apply(predict_vax_status, axis=1)

test.head()

```

	Unnamed: 0	vax_status2	WBscore	PCCsymp	predicted_vax_status
0	1	Vaccinated	Improved	Less	Vaccinated
1	2	Vaccinated	Improved	Less	Vaccinated
2	3	Vaccinated	Improved	Less	Vaccinated
3	4	Vaccinated	Worsened	Less	Vaccinated
4	5	Vaccinated	Unchanged	Less	Vaccinated

**1(g)**

```

from sklearn.naive_bayes import CategoricalNB
from sklearn.preprocessing import LabelEncoder

label_wb = LabelEncoder()
label_pcc = LabelEncoder()
label_y = LabelEncoder()

X_train = pd.DataFrame({
    "WBscore": label_wb.fit_transform(train["WBscore"]),
    "PCCsymp": label_pcc.fit_transform(train["PCCsymp"])
})

y_train = label_y.fit_transform(train["vax_status"])

X_test = pd.DataFrame({
    "WBscore": label_wb.transform(test["WBscore"]),
    "PCCsymp": label_pcc.transform(test["PCCsymp"])
})

model = CategoricalNB()
model.fit(X_train, y_train)

```

```
sklearn_preds = label_y.inverse_transform(model.predict(X_test))

# Compare with manual predictions
comparison = test["predicted_vax_status"].values == sklearn_preds

print("All predictions identical:", comparison.all())
```

All predictions identical: True

We see that running this code confirms a 100% match between the manual calculation and the Scikit-Learn implementation.