ECE 313 FINAL PRESENTATION

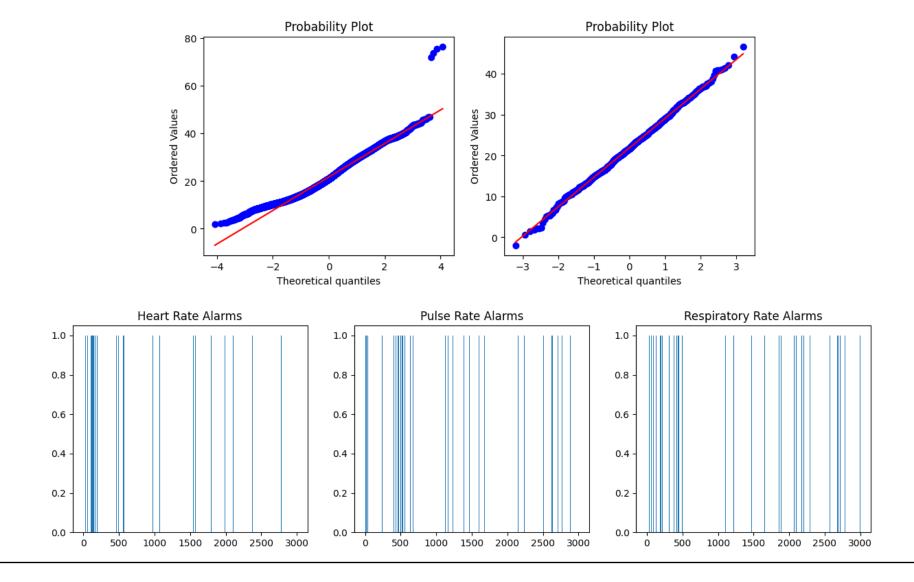
Hrishikesh Deshpande (hd11)

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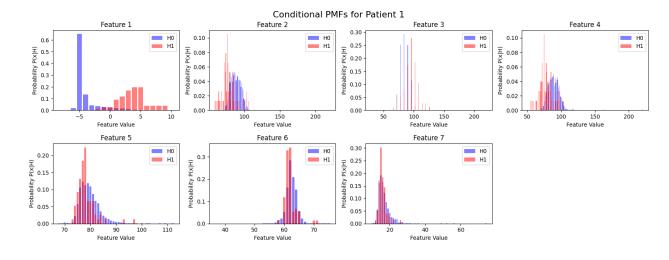
TASKS 1 & 2

- Generated PMFs, PDFs, and CDFs for respiratory rate samples (sizes 70, 1000, 30,000) and observed reduced fluctuations with larger sizes (hd11).
- Real patient data showed greater deviation from normality (RMSE of **0.9798**) than simulated data (RMSE of **0.2468**) and had a narrower right-skewed range (kappana2).
- Alarms were created using heart, pulse, and respiratory rate thresholds from empirical and normalized distributions (sg97).
- Normalized alarms had a lower error probability (**0.0160**) compared to empirical (**0.0327**), showing better detection performance (hd11).



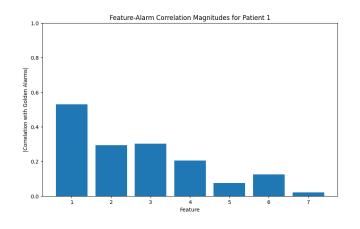
TASK 3

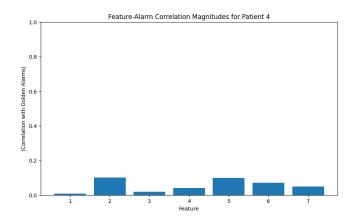
- Constructed conditional PMFs and likelihood matrices for each of 9 patients (sg97).
- Calculated prior probabilities P(H0) and P(H1) for each patient's training data (sg97).
- Built ML and MAP decision vectors from likelihoods and priors (kappana2).
- Used false alarm and miss detection rates to compute overall error probabilities (kappana2).

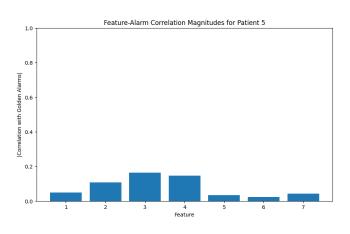


TASK 3 - PATIENT DECISIONS (HD11)

- Patient 1 (ML): Features 1 & 3 chosen due to both low error and high impact (0.0325 error).
- **Patient 4 (ML):** Features 2 & 5 chosen because feature 5 had lowest error and feature 2 provided strong signal correlation (**0.0332 error**).
- Patient 5 (MAP): Features 1 & 3 are the combined best from lowest-error and highest-impact pairs (0.0256 error).







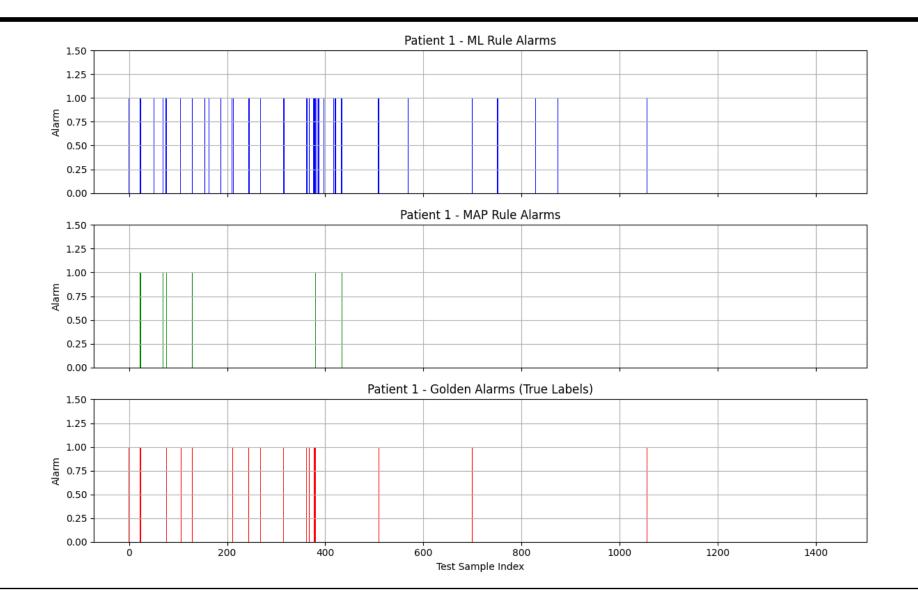
TASK 4

- Constructed joint likelihood matrices **P(X,Y|H0)** and **P(X,Y|H1)** for selected feature pairs (sg97):
 - Patient 1: Features 0 & 2
 - Patient 4: Features 1 & 4
 - Patient 5: Features 0 & 2
- Created joint ML and MAP decision rules using likelihood ratios and class priors (sg97).
- Generated predicted alarms for each patient and compared to golden labels (hd11).
- Error probabilities were computed for each method and patient (hd11).

| Patient | Method | P(FA) | P(MD) | P(E) |
|---------|--------|-------|-------|-------|
| Patient | ML | 3.98% | 0.14% | 2.06% |
| 1 | MAP | 0.42% | 1.33% | 0.45% |
| Patient | ML | 3.78% | 0% | 1.89% |
| 4 | MAP | 3.78% | 0% | 3.78% |
| Patient | ML | 3.28% | 0.07% | 1.67% |
| 5 | MAP | 0.35% | 0.07% | 0.35% |

TASK 4

- MAP rule outperformed ML in Patients 1 and 5 with lower total error (hd11):
 - Patient 1: MAP (0.45%) vs. ML (2.06%)
 - **Patient 5:** MAP (0.35%) vs. ML (1.67%)
- Patient 4 showed worse MAP performance, likely due to skewed priors misaligning with feature distribution (kappana2).
- Average Errors (kappana2):
 - ML Average Error: 1.87%
 - MAP Average Error: 1.53%
- MAP models are more sensitive to prior distribution assumptions but can dramatically reduce false alarms and misses when aligned well (kappana2).



BONUS TASK

- Built and evaluated **Logistic Regression** and **Neural Network** classifiers for alarm prediction (kappana2).
- Compared these to the **MAP Rule** from Task 4 using the same test set across all models (kappana2).
- Tested both models with (hd11):
 - T_ML (0.5) standard threshold
 - T_MAP threshold with class priors

| | Model | Threshold | P(False Alarm) | P(Miss Detection) | P(Error) |
|---|---------------------|-----------|----------------|-------------------|----------|
| 0 | Logistic Regression | τML (0.5) | 0.000086 | 0.010249 | 0.005168 |
| 1 | Logistic Regression | τΜΑΡ | 0.000000 | 0.010249 | 0.005124 |
| 2 | Neural Network | τML (0.5) | 0.000000 | 0.010249 | 0.005124 |
| 3 | Neural Network | τΜΑΡ | 0.000000 | 0.010249 | 0.005124 |
| 4 | MAP Rule (Task 4) | τΜΑΡ | 0.015174 | 0.004652 | 0.015265 |

BONUS TASK

- Both Logistic Regression and Neural Network significantly outperformed the MAP rule from Task 3 in total error rate (sg97).
- The MAP Rule had the lowest miss detection rate (0.46%) but suffered from a high false alarm rate (1.52%) (sg97).
- Neural Network and MAP Rule delivered the lowest total error (**0.51245**%) with zero false alarms (hd11).
- Incorporating priors via MAP Rule improved all models' decision boundaries without hurting recall (hd11).

QUESTIONS?