

pHase

Fermented Raw Milk Stage Prediction Model

CSI 5170/4170 - Machine Learning
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Introduction & Project Goals



Problem Definition

- Raw milk ferments naturally through predictable stages
- No ML model exists to estimate fermentation stage or pH
- Existing research focuses on pasteurized, lab-controlled dairy
- Consumers rely on guesswork (smell, texture)

Relevance to Machine Learning

- Nonlinear time-series modeling
- Regression + classification
- Real-world biological data
- Feature engineering & evaluation

Goals & Scope

- Predict pH from time, temperature, and culture conditions
- Predict discrete fermentation stages
- Build a real-time interactive prediction tool
- Scope: 12-jar dataset, 45 days, 67-69°F environment



Background & Literature

Existing Research Themes

- Yogurt/kefir studies: pasteurized milk, standard cultures
- Industrial fermentation is highly controlled
- Spoilage detection: fresh vs spoiled classification

Key Gaps

- Lack of raw milk datasets at household temperatures
- No ML models for full fermentation stages
- Rarely models natural microbial diversity
- Consumer prediction tools do not exist

pHase Contribution

- Real raw milk dataset
- Full stage modeling (fresh to cheese)
- Focus on natural fermentation, not industrial



Methodology Pt. 1

Study Design

- 12 jars of fresh raw milk
- Jars 1–6: no culture
- Jars 7–12: 1 tbsp naturally fermented milk added
- Stored 45 days at 67–69°F

Data Collection

- 50 measurements per jar -> 600 total entries
- pH measured together for consistency
- High-frequency sampling early in fermentation
- Slower sampling later as changes slowed

Recorded Variables

- Date, time, jar ID, pH, culture status



Methodology Pt. 2

Preprocessing

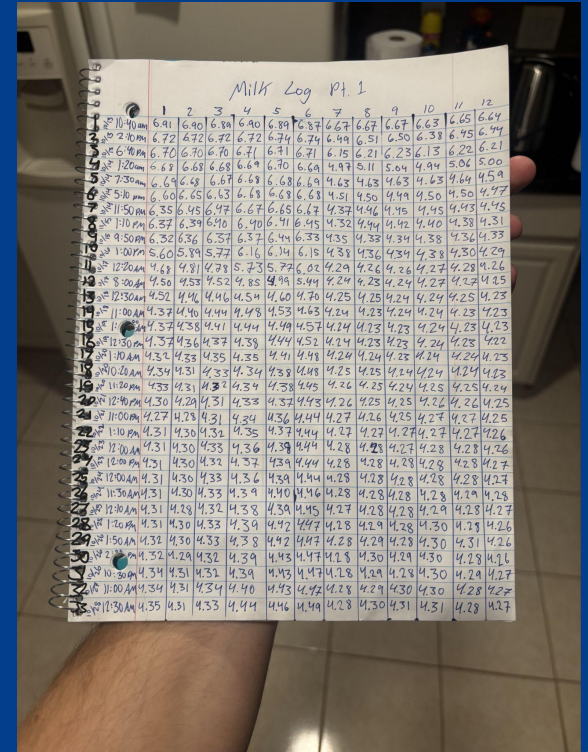
- Merge date + time -> timestamp
- Convert to hours since fermentation start
- Exponential smoothing to reduce noise
- Engineered features:
 - Rate of pH change, Time^2 , $\text{pH} \times \text{time}$

Stage Labeling

- Based on known pH ranges + biological order
- Stages forced to move forward only

Models Used

- Random Forest Regressor (pH)
- Random Forest Classifier (stage)
- Train/val/test split by jar and culture group (4/1/1)



	1	2	3	4	5	6	7	8	9	10	11	12
10:40 am	6.91	6.90	6.88	6.89	6.87	6.67	6.67	6.67	6.67	6.65	6.64	
11:20 am	6.72	6.72	6.72	6.74	6.74	6.49	6.51	6.50	6.38	6.45	6.44	
12:00 pm	6.70	6.70	6.71	6.71	6.71	6.15	6.21	6.23	6.13	6.22	6.21	
12:40 pm	6.68	6.68	6.69	6.69	6.70	6.69	6.11	6.09	6.49	5.06	5.06	
1:20 pm	6.64	6.64	6.64	6.68	6.68	6.69	6.43	6.43	6.63	4.63	4.64	
2:00 pm	6.60	6.65	6.63	6.69	6.69	6.63	4.51	4.50	4.49	4.50	4.50	
2:40 pm	6.55	6.45	6.47	6.67	6.65	6.47	4.37	4.46	4.46	4.45	4.43	
3:20 pm	6.57	6.39	6.40	6.40	6.41	6.45	4.32	4.49	4.49	4.40	4.38	
4:00 pm	6.32	6.36	6.37	6.37	6.44	6.33	4.35	4.33	4.34	4.38	4.36	
4:40 pm	6.40	5.89	5.77	6.16	6.14	6.18	4.38	4.36	4.39	4.38	4.30	
5:20 pm	4.41	4.41	4.78	5.73	5.77	6.02	4.29	4.26	4.26	4.27	4.28	
6:00 pm	4.50	4.53	4.52	4.85	4.99	5.94	4.24	4.23	4.24	4.27	4.27	
6:40 pm	4.52	4.46	4.46	4.54	4.60	4.70	4.25	4.25	4.24	4.24	4.25	
7:20 pm	4.37	4.40	4.44	4.48	4.53	4.63	4.24	4.23	4.24	4.24	4.23	
8:00 pm	4.37	4.38	4.41	4.44	4.49	4.57	4.14	4.13	4.14	4.13	4.13	
8:40 pm	4.37	4.36	4.37	4.38	4.44	4.52	4.14	4.13	4.13	4.13	4.13	
9:20 pm	4.34	4.33	4.35	4.35	4.41	4.49	4.14	4.14	4.14	4.14	4.14	
10:00 am	4.34	4.31	4.33	4.34	4.39	4.48	4.15	4.15	4.14	4.14	4.14	
10:40 am	4.33	4.31	4.33	4.34	4.39	4.48	4.15	4.15	4.14	4.14	4.14	
11:20 am	4.30	4.29	4.31	4.31	4.36	4.44	4.17	4.16	4.16	4.16	4.16	
12:00 pm	4.31	4.30	4.31	4.31	4.37	4.44	4.17	4.17	4.17	4.17	4.17	
12:40 pm	4.31	4.30	4.31	4.31	4.39	4.44	4.18	4.18	4.18	4.18	4.18	
1:20 pm	4.30	4.30	4.32	4.32	4.39	4.44	4.18	4.18	4.18	4.18	4.18	
2:00 pm	4.31	4.30	4.31	4.31	4.39	4.44	4.18	4.18	4.18	4.18	4.18	
2:40 pm	4.31	4.31	4.32	4.32	4.39	4.44	4.18	4.18	4.18	4.18	4.18	
3:20 pm	4.31	4.31	4.32	4.32	4.39	4.44	4.18	4.18	4.18	4.18	4.18	
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12:00 pm	4.31	4.31	4.32	4.32	4.39	4.44	4.18	4.18	4.18	4.18	4.18	

Evaluation: Research Question

Main Research Question:

- Can raw milk fermentation be predicted accurately using ML?

Specific Focus

- Can pH be modeled as a function of time with simple features?
- Can fermentation stage be predicted reliably?
- Is the process predictable enough for real time user tools?



Evaluation: Experimental Setup

Libraries & Tools

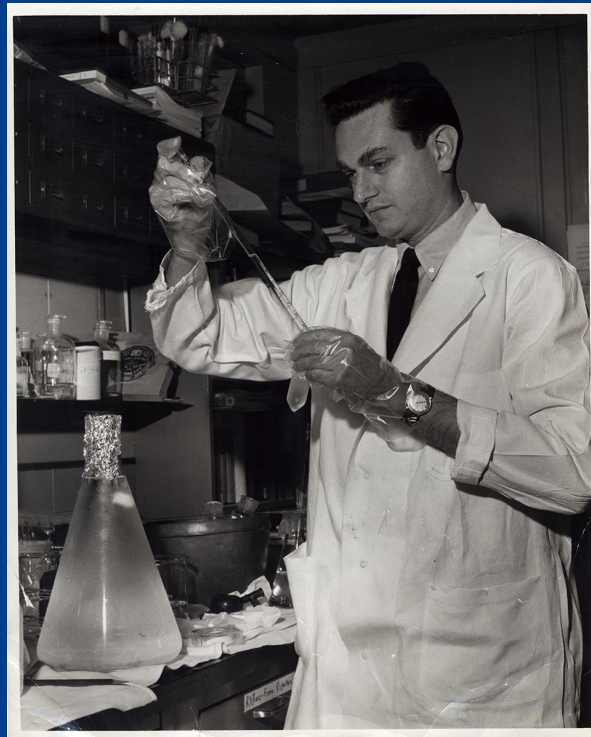
- Google Colab
- Scikit-learn
- Pandas / NumPy
- Matplotlib
- ZoneInfo, datetime
- ipywidgets (interactive tool)

Hyperparameters

- 400 trees for both RF models
- max_depth = None
- Grouped jar split (culture vs no culture)

Evaluation Data

- Predictions vs actual pH
- Stage predictions vs stage labels
- Error analysis and visual graphs



Evaluation: Results

Regression Model (pH)

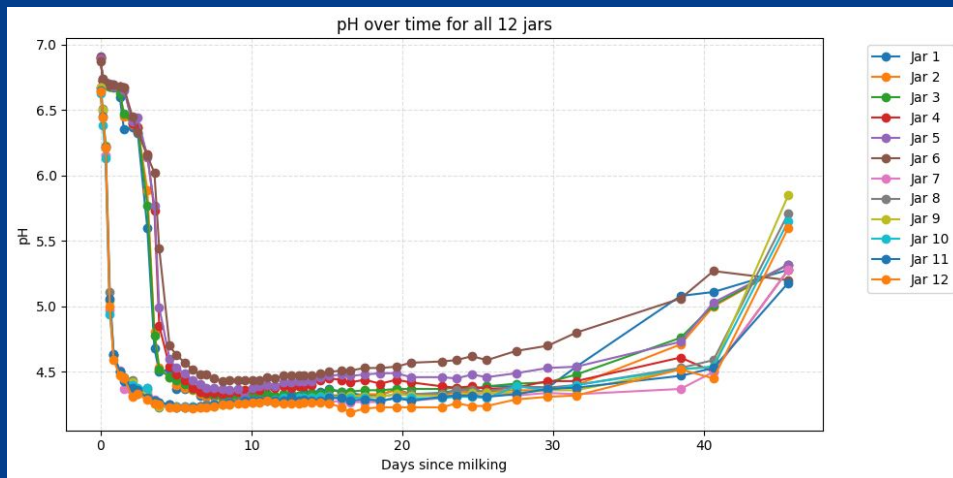
- RMSE ≈ 0.0375
- $R^2 \approx 0.9976$
- Closely tracks real pH curves

Classification Model (Stage)

- 100% test accuracy
- Perfect F1 score
- Correctly identified all stages

Observations

- Culture jars drop pH faster early
- Non-culture jars shift more slowly
- Interactive tool visualizes differences



Evaluation: Discussion

Key Findings

- Raw milk fermentation follows a predictable pattern
- Time + culture status explain most variation
- Stage labeling method worked well

Comparison to Literature

- Industrial studies don't model natural fermentation
- Spoilage models ignore intermediate stages
- pHase provides full-stage prediction

Additional Observations

- Jars lost ~20% volume during fermentation
- Culture adds significant speed early on

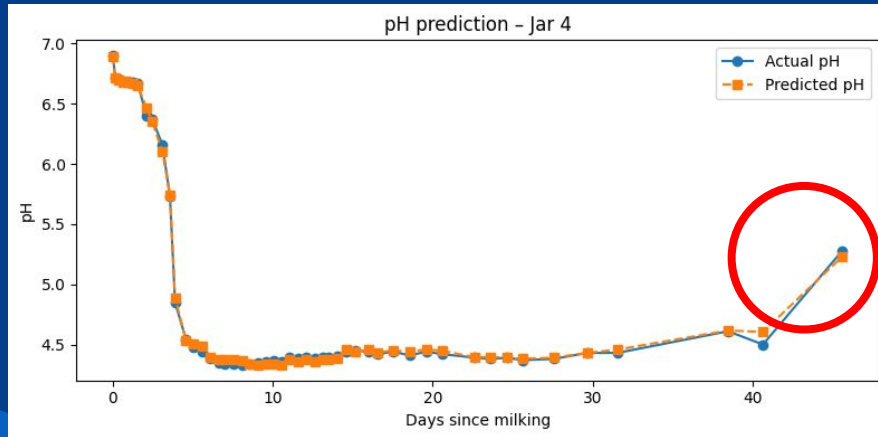




pHase - Demo

Limitations

- Limited to 12 jars, 1 milk source
- Only one temperature range (67–69°F)
- Late cheese stage data incomplete
- pH readings take 1–2 minutes per jar
- Time and resource constraints



Future Work & Conclusion

Conclusion

- ML can model raw milk fermentation accurately
- pH and stages follow stable patterns
- pHase provides real-time predictions

Future Work

- Use multiple farms and more jars
- Collect longer and more frequent measurements
- Improve stage labeling with direct observations
- Add bacterial testing
- Expand interactive tool into a public app



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Thanks!

Do you have any questions?
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