

The sample calculation of the CPLIP model in the benchmark stage

This is a **retrospective cohort study** of patients with oral potentially malignant disorders (OPMDs) with a **36-month** follow-up window from the index diagnostic biopsy/WSI. We use a CPLIP-based risk model to classify patients **at baseline** as high risk or low risk.

The outcome is malignant transformation to oral cancer within 36 months, confirmed by biopsy. For validation, the 36-month biopsy-confirmed outcome serves as the **ground truth**, and we evaluate performance at the **patient level**.

	No Progression	Progression
Low-Risk		
High-Risk		

Overall, the malignant transformation rate across all OPMD groups was 7.9% (99% confidence interval [CI], 4.9%-11.5%)_(Iocca et al., 2020). Splits are patient-level (no patient appears in more than one split).

How many patients to collect for the benchmark

Use the planning rule: patients = target events ÷ event rate

- To target ≈ 100 events (a solid benchmark for pooled 5-fold CV):
 $N = 100 / 0.079 = 1,265.8 \rightarrow \sim 1,300$ **patients** (expected ≈ 103 events).
This gives ≈ 21 events per fold in 5-fold CV.
- Add buffer for missing outcomes (e.g., **10% loss** to follow-up):
 $1,266 / 0.90 \approx \sim 1,400$ -1,450 patients.
- If I want to be conservative and plan against the lower bound of the rate (4.9%):
 $100 / 0.049 \approx 2,041 \rightarrow \sim 2,250$ if I also allow 10% loss.
- Minimal pilot (pipeline shake-down): ~ 650 patients (~ 50 events).
- Tighter AUC CIs ($\approx \pm 0.05$): $\sim 1,600$ –1,800 patients (≈ 125 -140 events).

- Above the cutoff → label High-risk
- Below the cutoff → label Low-risk

Assume **N=1,300 patients** and 7.9% will transform.

Choose a cutoff that gives Sensitivity ≥ 0.90

→ expect about **570 high-risk / 730 low-risk** ($\approx 44\%$ / 56%).

How to do this in 5-fold CV

In the 5-fold CV steps, the threshold is chosen inside the training data via inner validation (nested CV) to avoid leakage; then apply it to the held-out fold.

- ① Split patients into 5 folds (patient-level, stratified).
- ② For each fold, train on the other 4 folds.
- ③ Choose the cutoff inside the training data (e.g., Sensitivity ≥ 0.90 , then maximize Specificity).
- ④ Apply that cutoff to the held-out fold; record who is high/low.
- ⑤ Pool all 5 held-out predictions to report final counts and metrics.

5-fold CV is better than 70:15:15

- More test signal. With N=1,300 and 7.9% events:
70:15:15 → test has ~ 195 people → ~ 15 events.
5-fold CV → everyone is tested once → $\sim 1,300$ people → ~ 103 events.
More events = tighter, more trustworthy metrics.
- Better training. Each CV model trains on $\sim 80\%$ of the data (vs 70%), so it learns more.
- Less luck. CV averages over 5 different test folds, not one lucky/unlucky split.
- Practicality. To get ~ 100 test events with a 15% test set, I'd need $\sim 8,440$ total patients. CV gives ~ 100 test events now by pooling the 5 folds.

locca, O., Sollecito, T. P., Alawi, F., Weinstein, G. S., Newman, J. G., De Virgilio, A., Di Maio, P., Spriano, G., Pardiñas López, S., & Shanti, R. M. (2020). Potentially malignant disorders of the oral cavity and oral dysplasia: A systematic review and meta-analysis of malignant transformation rate by subtype. *Head & neck*, 42(3), 539-555.