Project: MontageAgent Author: Prem Kondru, IIT Guwahati

MontageAgent – Architecture & Rationale

Project: MontageAgent — AI Agent for Event Photo Curation & Instagram Publishing

Author: Prem Kondru (BTech Engineering Physics, IIT Guwahati; Photography Club: Montage)

Purpose & Problem

At IIT Guwahati's Photography Club (Montage), every event generates hundreds of photos. The manual workflow involves:

- 1. Removing duplicates
- 2. Grouping by theme/moment
- 3. Writing captions in a consistent style
- 4. Assembling Instagram carousel posts

This process is repetitive, error-prone, and consumes hours of human effort.

MontageAgent automates the workflow end-to-end:

 $Ingest \rightarrow Dedupe \rightarrow Categorize \rightarrow Cluster \rightarrow Caption \rightarrow Export/Publish$

Interaction Flow

- 1. **User Input:** Upload images, set labels, choose captioner mode, define event name, adjust max images per post.
- 2. **Pipeline Execution:** Ingest → Embed → Deduplicate → Categorize → Cluster → Caption → Export.
- 3. Captioning:
 - o **BLIP-2 mode:** Generate per-image captions, extract common words, produce abstract caption.
 - o Template mode: Deterministic, rule-based caption.
 - o Hashtags pulled from base + labels + historical RAG hints.
- 4. Preview & Export: IG-style preview, per-image inclusion/exclusion, JSON export for carousel.

How the Agent Works

Reasoning

- **Visual Understanding:** CLIP embeddings assess similarity, remove near-duplicates, and infer labels (zero-shot).
- Captioning: Two captioner modes:
 - o **Template Mode** → Deterministic, rule-based captions.
 - \circ BLIP-2 (with LoRA) \rightarrow Generates captions based on a set of cluster images.
- **Stylistic Consistency:** Retrieval-augmented input (RAG) over past captions maintains tone and ensures continuity across posts.

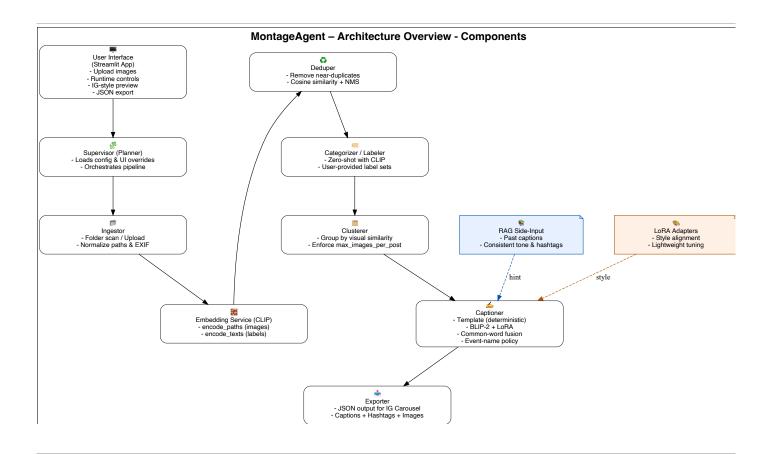
Planning

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- A Supervisor (Planner) module orchestrates a fixed pipeline:
 Ingest → Embed → Dedupe → Categorize → Cluster → Caption → Export.
- The plan is deterministic, simplifying debugging and evaluation.

Execution

- Executors (Workers): Each step runs as an independent tool with clear inputs/outputs.
- **Human-in-the-loop UI:** Built with Streamlit, allowing users to include/exclude photos, adjust labels, and preview posts before export.



Components

- **User Interface:** Streamlit app with IG-style previews, zoom, toggles, runtime configs, and JSON export.
- Supervisor (Planner): Loads config, applies UI overrides, sequences executors.
- **Ingestor:** Scans folders/uploads, builds image list.
- Embedding Service (CLIP): Provides encode_paths and encode_texts for similarity, clustering, and label assignment.
- **Deduper:** Removes near-duplicates via cosine similarity.
- Categorizer/Labeler: Zero-shot label assignment using CLIP and user-provided label sets.
- Clusterer: Groups images by visual similarity; enforces max_images_per_post with balanced sampling.
- Captioner:
 - o **Template mode:** Deterministic phrasing.

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- o **BLIP-2 mode:** Batch captioning + LoRA fine-tuning; fuses outputs into one abstract caption. Supports event-name injection policies (off | hint | only proper noun).
- Exporter: Builds JSON for Instagram Posts upload: {caption, hashtags, images[]}.

Key Design Choices & Agent Patterns

https://www.anthropic.com/engineering/building-effective-agents https://arxiv.org/pdf/2405.1046

- Role-based cooperation (multi-agent): Supervisor + tool workers (ingest, embed, dedupe, cluster, caption, export).
- **Fixed pipeline:** Single path improves reliability, reproducibility, and transparency.
- **Prompt/response optimisation:** strict caption prompts; output guards (no hashtags, abstract, proper-noun policy).
- CLIP for structural tasks (dedupe, clustering, labels): Efficient, lightweight, and non-generative.
- RAG: style/hashtag hints from nearest past captions.
- BLIP-2 + LoRA for style: Strong caption quality with minimal compute/storage overhead.
- Config-driven + UI overrides: Enables repeatability, A/B testing, and flexible runtime control.
- **Guardrails:** Caption length limits, "event name only" injection, and exclusion of proper nouns ensure alignment with club style.
- Evaluator: CLIPScore & silhouette surfaced to the user; iterative feedback loop.

Models & Rationale

- OpenCLIP (model ViT-B/32 pretrained laion2b s34b b79k):
 - o Lightweight, efficient visual similarity.
 - o Ideal for deduplication, clustering, zero-shot labels, CLIPScore.
- RAG over Past Captions:
 - o Ensures consistency in tone/hashtags.
 - o Avoids over-fitting to rigid templates.
- BLIP-2 (base model Salesforce/blip2-flan-t5-xl) with LoRA:
 - o Generates abstract, mood-driven captions aligned with club style.
 - o LoRA enables style specialization without full finetuning.
 - o Per-image batching + common-word fusion prioritizes shared visual cues.
 - o Guardrails enforce reliability and stylistic alignment.

Evaluation: Quality & Reliability Metrics

- CLIPScore: Measures caption—image alignment (per-image and cluster means).
- Silhouette Score: Quick proxy for clustering cohesion/separation.
- **Dedupe Rate:** % of near-duplicates removed.
- Human Ratings (optional): Abstractness, tone, Instagram readiness.
- A/B Protocol: Compare template vs BLIP-2/LoRA on the same clusters; track metrics per event and overall.

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• A full **Data Science Report (PDF)** and **Interaction Logs** (prompts + chat history) are generated to document method and outcomes.

<u>In summary, Montage is a complete, AI agent that automates a real university workflow, integrates a LoRA-tuned model for style-safe captioning, and ships with the metrics and documentation required to evaluate and maintain it.</u>