

AI-Powered Slope Stability Monitoring System

Complete Pipeline from Drone Images to FOS Calculation

PHASE 1: Data Acquisition & Preprocessing

1.1 Drone Image Capture Strategy

| pyt | hon |
|---------|---|
| Flig | Optimal flight parameters for slope monitoring ght Planning: — Overlap: 80% front, 70% side overlap — Resolution: <2cm/pixel for crack detection — Flight pattern: Grid + oblique angles (45°) |
| | — Multiple flights: Different seasons/conditions — GPS accuracy: RTK/PPK for precise georeferencing Camera settings |
| - - | RAW format for maximum detail Fixed exposure for consistent lighting Multiple angles: Nadir + oblique views Stereo pairs for 3D reconstruction |

1.2 Data Types to Collect

- RGB Images: Visual crack detection
- LiDAR Data: Precise 3D geometry (if available)
- Multispectral: Vegetation health, moisture detection
- Thermal: Temperature variations, seepage detection
- **GPS Coordinates**: Precise georeferencing

PHASE 2: 3D Model Construction

2.1 Photogrammetry Pipeline

A) Structure from Motion (SfM)

| python | |
|--------|--|
| | |
| | |

```
# Software options:

Commercial:

— Agisoft Metashape (Best accuracy)

— Pix4D (Industry standard)

— RealityCapture (Fast processing)

Open Source:

— OpenDroneMap (Free, good quality)

— COLMAP (Research-grade)

— MeshRoom (Alembic ecosystem)

# Process flow:

Images → Feature matching → Camera poses → Dense point cloud → 3D mesh
```

B) Point Cloud Processing

```
# Key steps:

1. Noise removal — Remove outliers, artifacts

2. Classification → Ground, vegetation, rock faces

3. Meshing — Convert points to 3D surfaces

4. Texture mapping ► Apply original images to 3D model

5. Georeferencing → Align to real-world coordinates

# Tools:

| CloudCompare (Free, powerful)
| PCL (Programming library)
| Open3D (Python-based)
| MeshLab (Mesh processing)
```

2.2 Quality Assessment

```
# Check 3D model quality:

Point density — → > 100 points/m² for crack detection

Accuracy — → <5cm error vs ground truth

Completeness — → No holes in critical areas

Texture quality — ➤ Sharp, well-aligned textures
```

3.1 Multi-Target Detection System

A) Crack Detection

| python |
|---|
| # Best models for slope cracks: |
| Primary: SAM (Segment Anything) |
| — Zero-shot crack segmentation |
| Works on 3D texture maps |
| Highly accurate for linear features |
| —— Fighty accurate for linear features |
| Alternative: EfficientNet-UNet |
| —— 95%+ accuracy on crack datasets |
| Fine-tuned for geological features |
| —— Pixel-level crack mapping |
| . Decree and a decree property |
| # Implementation: |
| from segment_anything import SamPredictor, sam_model_registry |
| import cv2 |
| |
| def detect_cracks_on_3d_model(texture_image): |
| sam = sam_model_registry["vit_h"](checkpoint="sam_vit_h.pth") |
| predictor = SamPredictor(sam) |
| |
| # Process each texture patch |
| predictor.set_image(texture_image) |
| # Use automatic mask generation for full coverage |
| masks = predictor.generate_masks() |
| |
| return crack_masks |
| |

B) Motion Detection (Displacement Analysis)

| python | | | |
|--------|--|--|--|
| | | | |
| | | | |
| | | | |

| # Multi-temporal analysis: Time Series Analysis: Compare 3D models from different dates Calculate displacement vectors Identify movement patterns Detect acceleration/deceleration |
|---|
| # Tools & Methods: CloudCompare M3C2: — Precise distance calculations between point clouds — Statistical significance testing — Uncertainty quantification — 3D displacement vectors |
| # Python implementation: import open3d as o3d import numpy as np |
| <pre>def detect_motion(cloud_t1, cloud_t2, threshold=0.05): # Register point clouds transformation = register_point_clouds(cloud_t1, cloud_t2)</pre> |
| # Calculate distances distances = np.asarray(cloud_t1.compute_point_cloud_distance(cloud_t2)) |
| # Identify significant movements moving_points = distances > threshold return moving_points, distances |

C) Joint Set Analysis

| python | | |
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| | | |
| | | |
| | | |
| | | |

```
# Discontinuity detection:
Automated Joint Analysis:
—— Plane fitting algorithms
  — Orientation analysis (dip/dip direction)

    Spacing measurements

--- Persistence calculations
L— Joint set classification
# Implementation approach:
def analyze_joint_sets(point_cloud):
  # 1. Segment planar surfaces
  planes = segment_planes_ransac(point_cloud)
  # 2. Calculate orientations
  orientations = []
  for plane in planes:
    dip, dip_direction = calculate_orientation(plane.normal)
    orientations.append((dip, dip_direction))
  # 3. Cluster similar orientations
  joint_sets = cluster_orientations(orientations)
  # 4. Calculate geometric properties
  for joint_set in joint_sets:
    spacing = calculate_spacing(joint_set)
    persistence = calculate_persistence(joint_set)
    roughness = calculate_roughness(joint_set)
  return joint_sets
```

3.2 Integration with 3D Models

| python | | |
|--------|--|--|
| | | |
| | | |
| | | |

```
# Map 2D detections to 3D coordinates:

def map_2d_to_3d(detection_mask, texture_coordinates, point_cloud):

# Convert 2D pixel coordinates to 3D world coordinates

world_coords = []

for pixel in detection_mask:

if pixel in texture_coordinates:

world_coord = texture_coordinates[pixel]

world_coords.append(world_coord)

return world_coords
```

III PHASE 4: Geotechnical Parameter Extraction

4.1 Rock Mass Classification Parameters

A) From AI Detection Results

```
python
# Extract geotechnical parameters:
def extract_geotechnical_params(cracks, joints, motion_data):
  parameters = {}
  # RQD (Rock Quality Designation)
  parameters['RQD'] = calculate_rqd_from_cracks(cracks)
  # Joint parameters
  parameters['joint_spacing'] = calculate_joint_spacing(joints)
  parameters['joint_orientation'] = get_joint_orientations(joints)
  parameters['joint_persistence'] = calculate_persistence(joints)
  parameters['joint_roughness'] = estimate_roughness(joints)
  parameters['joint_aperture'] = measure_aperture(cracks)
  # Weathering assessment
  parameters['weathering_grade'] = assess_weathering_from_texture()
  # Groundwater conditions
  parameters['groundwater'] = detect_seepage_signs()
  return parameters
```

B) Rock Mass Rating (RMR) Calculation

```
python
def calculate_rmr(parameters):
  rmr_score = 0
  # 1. Uniaxial compressive strength (from rock type/field tests)
  rmr_score += get_ucs_rating(parameters['rock_type'])
  # 2. RQD
  rmr_score += get_rqd_rating(parameters['RQD'])
  # 3. Joint spacing
  rmr_score += get_spacing_rating(parameters['joint_spacing'])
  # 4. Joint condition
  rmr_score += get_joint_condition_rating(
    parameters['joint_persistence'],
    parameters['joint_roughness'],
    parameters['joint_aperture']
  # 5. Groundwater conditions
  rmr_score += get_groundwater_rating(parameters['groundwater'])
  return rmr_score
```

4.2 Strength Parameters

```
# Convert RMR to strength parameters:

def rmr_to_strength_params(rmr_score):

# Hoek-Brown parameters

mb = np.exp((rmr_score - 100) / 28)

s = np.exp((rmr_score - 100) / 9)

a = 0.5 + (np.exp(-rmr_score/15) - np.exp(-20/3)) / 6

# Mohr-Coulomb parameters

friction_angle = 20 + 30 * (rmr_score / 100)

cohesion = 0.05 + 0.25 * (rmr_score / 100) # MPa

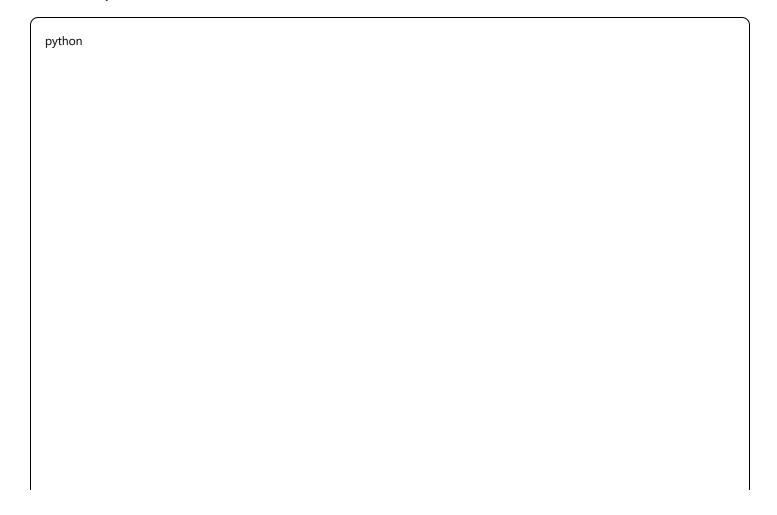
return {

    'friction_angle': friction_angle,
    'cohesion': cohesion,
    'mb': mb, 's': s, 'a': a
}
```

PHASE 5: Factor of Safety (FOS) Calculation

5.1 Slope Stability Analysis Methods

A) Limit Equilibrium Methods



```
# Bishop's Simplified Method
def bishops_method(slope_geometry, soil_params, water_table):
  slices = discretize_slope(slope_geometry)
  for iteration in range(max_iterations):
     F = 0 # Factor of safety
     for slice in slices:
       # Calculate forces on each slice
       W = slice.weight
       u = calculate_pore_pressure(slice, water_table)
       # Normal and shear forces
       N = W * np.cos(slice.alpha) + (slice.X_right - slice.X_left)
       T = W * np.sin(slice.alpha) + (slice.E_right - slice.E_left)
       # Available shear strength
       s_available = (slice.c * slice.width +
                (N - u * slice.width) * np.tan(slice.phi))
       F += s_available / T
     if converged(F):
       break
  return F
# Janbu's Method (for non-circular surfaces)
def janbu_method(slope_geometry, soil_params):
  # Similar implementation for irregular failure surfaces
  pass
```

B) Finite Element Method (Advanced)

```
# For complex geometries and stress analysis
import numpy as np
from scipy.sparse import csc_matrix
from scipy.sparse.linalg import spsolve

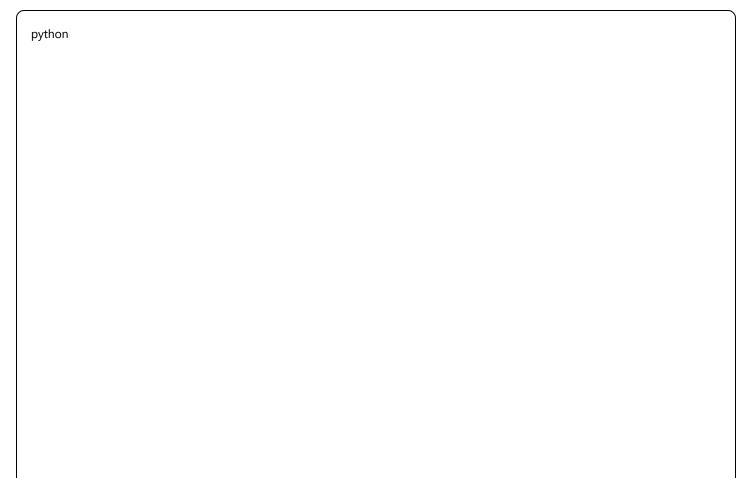
def fem_slope_analysis(mesh, material_properties, boundary_conditions):
    # Assemble stiffness matrix
    K = assemble_stiffness_matrix(mesh, material_properties)

# Apply boundary conditions
K, F = apply_boundary_conditions(K, boundary_conditions)

# Solve for displacements
displacements = spsolve(K, F)

# Calculate stresses and safety factors
stresses = calculate_stresses(displacements, material_properties)
local_fos = calculate_local_safety_factors(stresses, material_properties)
return np.min(local_fos) # Critical FOS
```

5.2 3D Slope Stability



```
def calculate_3d_fos(point_cloud, joint_sets, strength_params):
    # Identify potential failure surfaces
failure_surfaces = identify_failure_surfaces(joint_sets)

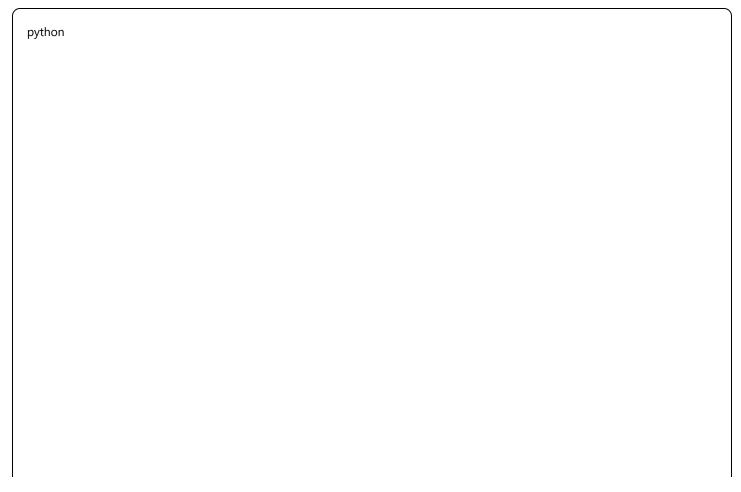
fos_values = []
for surface in failure_surfaces:
    # Extract geometry along failure surface
    surface_geometry = extract_surface_geometry(point_cloud, surface)

# Calculate 3D limit equilibrium
fos = calculate_3d_limit_equilibrium(
    surface_geometry,
    strength_params,
    joint_orientations=surface.orientations
)
fos_values.append(fos)

return min(fos_values), fos_values
```

PHASE 6: 2D Stability Mapping

6.1 Create Stability Maps



```
import matplotlib.pyplot as plt
import geopandas as gpd
from shapely.geometry import Point, Polygon
def create_stability_map(slope_area, fos_values, coordinates):
  # Create grid for interpolation
  grid_x, grid_y = np.meshgrid(
     np.linspace(coordinates.x.min(), coordinates.x.max(), 100),
     np.linspace(coordinates.y.min(), coordinates.y.max(), 100)
  # Interpolate FOS values
  from scipy.interpolate import griddata
  fos_grid = griddata(
     (coordinates.x, coordinates.y),
     fos_values,
     (grid_x, grid_y),
     method='cubic'
  )
  # Create stability classification
  stability_classes = np.zeros_like(fos_grid)
  stability_classes[fos_grid > 1.5] = 4 # Safe
  stability_classes[(fos_grid > 1.2) & (fos_grid <= 1.5)] = 3 # Good
  stability_classes[(fos_grid > 1.0) & (fos_grid <= 1.2)] = 2 # Caution
  stability_classes[fos_grid <= 1.0] = 1 # Critical
  return grid_x, grid_y, stability_classes
```

6.2 Visualization

```
def plot_stability_map(grid_x, grid_y, stability_classes, detected_features):
  fig, ax = plt.subplots(1, 1, figsize=(12, 8))
  # Color map for stability zones
  colors = ['red', 'orange', 'yellow', 'green']
  im = ax.contourf(grid_x, grid_y, stability_classes,
             levels=4, colors=colors, alpha=0.7)
  # Overlay detected features
  for crack in detected_features['cracks']:
     ax.plot(crack.x, crack.y, 'k-', linewidth=2, label='Cracks')
  for joint in detected_features['joints']:
     ax.plot(joint.x, joint.y, 'b--', linewidth=1, label='Joints')
  # Add legend and labels
  ax.set_xlabel('Easting (m)')
  ax.set_ylabel('Northing (m)')
  ax.set_title('Slope Stability Map')
  # Color bar
  cbar = plt.colorbar(im, ax=ax)
  cbar.set_label('Stability Class')
  cbar.set_ticks([1, 2, 3, 4])
  cbar.set_ticklabels(['Critical', 'Caution', 'Good', 'Safe'])
  return fig, ax
```

PHASE 7: Alert System & Monitoring

7.1 Risk Assessment

```
def assess_risk_level(fos, crack_density, motion_rate):
  risk_score = 0
  # FOS contribution (50% weight)
  if fos < 1.0:
     risk_score += 50
  elif fos < 1.2:
    risk_score += 30
  elif fos < 1.5:
     risk_score += 15
  # Crack density contribution (30% weight)
  if crack_density > 0.8: # cracks per m<sup>2</sup>
     risk_score += 30
  elif crack_density > 0.5:
    risk_score += 20
  elif crack_density > 0.2:
     risk_score += 10
  # Motion rate contribution (20% weight)
  if motion_rate > 10: # mm/year
     risk_score += 20
  elif motion_rate > 5:
    risk_score += 15
  elif motion_rate > 2:
     risk_score += 10
  # Classify risk level
  if risk_score >= 70:
     return "CRITICAL"
  elif risk_score > = 50:
    return "HIGH"
  elif risk_score > = 30:
     return "MODERATE"
  else:
     return "LOW"
```

7.2 Automated Monitoring

```
class SlopeMonitoringSystem:
  def __init__(self):
     self.baseline model = None
     self.alert_thresholds = {
       'fos_critical': 1.0,
       'fos_warning': 1.2,
       'motion_critical': 10, # mm/year
       'new_crack_threshold': 0.1 # m length
  def process_new_survey(self, new_images, new_lidar=None):
     # 1. Build 3D model
     new_model = build_3d_model(new_images, new_lidar)
     # 2. Detect features
     cracks = detect_cracks(new_model)
     joints = analyze_joint_sets(new_model)
     motion = detect_motion(self.baseline_model, new_model)
     # 3. Calculate FOS
     fos_map = calculate_fos_distribution(new_model, cracks, joints)
     # 4. Assess changes
     alerts = self.check_for_alerts(cracks, motion, fos_map)
     # 5. Update baseline
     self.baseline_model = new_model
     return {
       'stability_map': fos_map,
       'detected_features': {'cracks': cracks, 'joints': joints},
       'motion_analysis': motion,
       'alerts': alerts
  def check_for_alerts(self, cracks, motion, fos_map):
     alerts = []
     # Check FOS values
     critical_fos_areas = np.where(fos_map < self.alert_thresholds['fos_critical'])</pre>
     if len(critical_fos_areas[0]) > 0:
       alerts.append({
          'type': 'CRITICAL_FOS',
```

```
'severity': 'HIGH',
    'locations': critical_fos_areas,
    'message': 'Factor of Safety below 1.0 detected'
  })
# Check for new cracks
new_major_cracks = [c for c in cracks if c.length > self.alert_thresholds['new_crack_threshold']]
if new_major_cracks:
  alerts.append({
    'type': 'NEW_CRACKS',
    'severity': 'MEDIUM',
    'count': len(new_major_cracks),
    'message': f'{len(new_major_cracks)} new significant cracks detected'
  })
# Check motion rates
high_motion_areas = np.where(motion.annual_rate > self.alert_thresholds['motion_critical'])
if len(high_motion_areas[0]) > 0:
  alerts.append({
    'type': 'EXCESSIVE_MOTION',
    'severity': 'HIGH',
    'locations': high_motion_areas,
    'message': 'Excessive ground movement detected'
  })
return alerts
```

PHASE 8: Complete Integration Pipeline

8.1 End-to-End Workflow

```
def complete_slope_monitoring_pipeline(drone_images, previous_model=None):
  """Complete pipeline from drone images to FOS calculation"""
  # Phase 1: 3D Reconstruction
  print("Building 3D model...")
  model_3d = build_photogrammetric_model(drone_images)
  # Phase 2: AI Feature Detection
  print("Detecting features with Al...")
  cracks = detect_cracks_sam(model_3d.texture_images)
  joints = analyze_joint_sets_3d(model_3d.point_cloud)
  if previous_model:
    motion = detect_motion_multitemporal(previous_model, model_3d)
  else:
    motion = None
  # Phase 3: Geotechnical Parameter Extraction
  print("Extracting geotechnical parameters...")
  geo_params = extract_geotechnical_parameters(cracks, joints, motion)
  strength_params = calculate_strength_parameters(geo_params)
  # Phase 4: FOS Calculation
  print("Calculating Factor of Safety...")
  fos_distribution = calculate_3d_fos_distribution(
    model_3d, joints, strength_params
  )
  # Phase 5: Create Stability Map
  print("Generating stability map...")
  stability_map = create_2d_stability_map(
    model_3d.coordinates, fos_distribution
  )
  # Phase 6: Risk Assessment & Alerts
  risk_assessment = assess_overall_risk(
    fos_distribution, cracks, motion
  return {
    '3d_model': model_3d,
    'detected_features': {
       'cracks': cracks,
```

```
'joints': joints,

'motion': motion

},

'geotechnical_params': geo_params,

'fos_distribution': fos_distribution,

'stability_map': stability_map,

'risk_assessment': risk_assessment

}
```

This complete pipeline takes you from raw drone images all the way to actionable slope stability insights with automated FOS calculations and risk mapping!