

Fallback spam detection documentation

Purpose and activation conditions

The fallback system guarantees lead processing continues if AI spam detection fails. It automatically activates when the AI node encounters:

- **API downtime:** Service outages or 5xx errors.
- **Rate limits:** 429 errors.
- **Auth issues:** Invalid/expired credentials.
- **Network failures:** Timeouts or unreachable endpoints.
- **Malformed responses:** Unexpected schema or parse errors.

When triggered, the workflow switches to deterministic, rule-based checks to classify leads without external dependencies.

- **Key property:** Both AI and fallback paths return the same output schema, allowing seamless merge and downstream routing.

Step-by-step nodes

Error trigger node

- **Type:** Error Trigger
- **Role:** Captures any failure from the AI Agent node and preserves the original lead payload.
- **Example payload captured:**

```
{ "error": { "message": "OpenAI API error: Rate limit exceeded", "name": "NodeApiError", "node": { "name": "Spam Detection AI Agent", "type": "n8n-nodes-base.agent" } }, "json": { "name": "John Doe", "email": "john@example.com", "phone": "555-1234", "message": "Interested in services" } }
```

- **Why it matters:** Original lead data is intact, enabling fallback processing without data loss.

Fallback detection node

- **Type:** Code (JavaScript)
- **Role:** Applies five categories of rules to determine spam vs. new lead.
- **Classification threshold:** Two or more indicators → “Possible Spam”; otherwise → “New Lead”.
- **Output schema:** Matches AI path:

- status
- reason
- spamIndicatorCount
- indicators[]
- all original lead fields

Full code (with comments)

```
// Get lead data from the previous node

const leadData = $input.item.json;

// Extract fields with fallbacks

const name = leadData.name || "";
const email = leadData.email || "";
const phone = leadData.phone || "";
const message = leadData.message || "";

// Initialize spam indicator counters

let spamIndicators = [];

// 1) EMAIL SPAM DETECTION

const emailLower = email.toLowerCase();
const emailTestKeywords = ["test", "spam", "fake", "example", "noreply"];
const hasEmailTestKeyword = emailTestKeywords.some(k => emailLower.includes(k));
const disposableDomains = [
  'tempmail', 'guerrillamail', '10minutemail', 'throwaway',
  'mailinator', 'trashmail', 'yopmail', 'temp-mail'
];
const isDisposableEmail = disposableDomains.some(d => emailLower.includes(d));
const isMissingAtSymbol = !email.includes('@');
```

```
const isEmailTooShort = email.length < 5;

if (hasEmailTestKeyword || isDisposableEmail || isMissingAtSymbol || isEmailTooShort) {

    spamIndicators.push('suspicious email');

}
```

// 2) NAME SPAM DETECTION

```
const nameLower = name.toLowerCase();

const testNames = ['test', 'asdf', 'qwerty', 'admin', 'user', 'demo'];

const isTestName = testNames.some(n => nameLower.includes(n));

const isNameTooShort = name.length < 2;

const isAllNumbers = /^d+$/ .test(name); // all digits

const isRepeatedChars = /^(.)\1+$/.test(name); // same char repeated (e.g., 'aaaa')

const isSingleCharRepeated = name.replace(/\s/g, "").length === 1;

if (isTestName || isNameTooShort || isAllNumbers || isRepeatedChars ||
isSingleCharRepeated) {

    spamIndicators.push('suspicious name');

}
```

// 3) PHONE SPAM DETECTION

```
const phoneDigits = phone.replace(/\D/g, ""); // strip non-digits

const fakePhonePatterns = [

    '5555555', '0000000', '1234567', '9999999',
    '1111111', '2222222', '3333333', '4444444',
    '6666666', '7777777', '8888888'

];

const isFakePhonePattern = fakePhonePatterns.some(p => phoneDigits.includes(p));

const isAllSameDigit = /^(\d)\1+$/.test(phoneDigits);
```

```
const isPhoneTooShort = phoneDigits.length < 7;
const isPhoneTooLong = phoneDigits.length > 15;
if (isFakePhonePattern || isAllSameDigit || isPhoneTooShort || isPhoneTooLong) {
    spamIndicators.push('suspicious phone');
}

// 4) MESSAGE SPAM DETECTION
const messageLower = message.toLowerCase();
const isMessageTooShort = message.length < 10;
const spamKeywords = [
    'viagra', 'casino', 'lottery', 'winner', 'congratulations',
    'click here', 'buy now', 'limited time', 'act now', 'free money',
    'nigerian prince', 'inheritance', 'bitcoin', 'crypto investment'
];
const hasSpamKeyword = spamKeywords.some(k => messageLower.includes(k));
const isAllCaps = message === message.toUpperCase() && message.length > 5;
const hasExcessivePunctuation = /[!?]{3,}/.test(message);
const urlCount = (message.match(/https?:\/\/[^\s]+/g) || []).length;
const hasTooManyUrls = urlCount > 2;
const isTestMessage = ['test', 'testing', 'asdf', 'hello'].includes(messageLower.trim());
if (isMessageTooShort || hasSpamKeyword || isAllCaps || hasExcessivePunctuation ||
hasTooManyUrls || isTestMessage) {
    spamIndicators.push('suspicious message');
}

// 5) MISSING DATA DETECTION
const requiredFields = [name, email, phone, message];
```

```

const missingFieldsCount = requiredFields.filter(f => !f || f === 'Not provided' || f.trim() === '').length;

if (missingFieldsCount >= 3) {
    spamIndicators.push(` ${missingFieldsCount} required fields missing`);
}

// FINAL CLASSIFICATION (threshold = 2)

const isSpam = spamIndicators.length >= 2;

let reason;

if (isSpam) {
    reason = `Multiple spam indicators detected (fallback rules): ${spamIndicators.join(', ')}`;
} else if (spamIndicators.length === 1) {
    reason = `Minor concern detected (fallback rules): ${spamIndicators[0]}, but overall appears legitimate`;
} else {
    reason = 'Passed basic validation (fallback rules)';
}

return {
    status: isSpam ? 'Possible Spam' : 'New Lead',
    reason,
    spamIndicatorCount: spamIndicators.length,
    indicators: spamIndicators,
    ...leadData
};

```

Code logic breakdown

1. Email detection logic

- **Test keywords:** Flags addresses containing “test”, “fake”, “example”, “spam”, “noreply”.
 - Examples: “test@example.com” → spam; “john@company.com” → clean
- **Disposable domains:** Flags common throwaway providers (tempmail, mailinator, yopmail, etc.).
 - Examples: “user@tempmail.com” → spam; “user@gmail.com” → clean
- **Invalid format:** Flags missing “@” or too-short addresses.
 - Examples: “notanemail” → spam; “a@b” → spam
- **Regex:** None; uses string checks and length validation.

2. Name detection logic

- **Test names:** Flags names containing “test”, “asdf”, “qwerty”, etc.
 - Examples: “Test User” → spam; “John Smith” → clean
- **Too short:** Less than 2 characters.
 - Examples: “A” → spam; “Jo” → clean
- **All numbers:** Entire string is digits.
 - Examples: “12345” → spam; “John123” → clean
- **Repeated chars:** Same character repeated (e.g., “aaaa”).
 - Examples: “aaaa” → spam; “Anna” → clean
- **Regex used:**
- **Digits:** `/^\d+$/`
- **Repeats:** `/(.)\1+$/`

3. Phone detection logic

- **Fake patterns:** Common placeholders (e.g., 555-5555).
 - Examples: “555-5555” → spam; “415-555-0198” → clean
- **All same digit:** Repeated single digit across the number.
 - Examples: “111-1111” → spam; “415-123-4567” → clean
- **Length bounds:** Fewer than 7 or more than 15 digits after stripping non-digits.
 - Examples: “123” → spam; “12345678901234567” → spam; “+1-415-555-0198” → clean
- **Regex used:**
- **Strip non-digits:** `\D/g`
- **All same digit:** `^(.)\1+$/`

4. Message detection logic

- **Too short:** Under 10 characters.
 - Examples: “hi” → spam; “I am interested in your services” → clean
- **Spam keywords:** Lottery, casino, buy now, etc.
 - Examples: “You won the lottery!” → spam; “I need pricing information” → clean
- **All caps:** Entire message uppercase and length > 5.
 - Examples: “BUY NOW!!!” → spam; “Please contact me” → clean
- **Excess punctuation:** 3+ consecutive “!” or “?”.
 - Examples: “Hello!!!” → spam; “Hello!” → clean
- **Excess URLs:** More than 2 links.

- **Regex used:**
- **Punctuation:** /[!?]{3,}/
- **URLs:** /https?:\/\/g

5. Missing data logic

- **Rule:** Count missing/“Not provided”/empty fields among name, email, phone, message.
- **Threshold:** 3 or more missing → one spam indicator.
- Examples: Three missing fields → spam indicator; one missing field → clean.

Classification threshold rationale

- **Threshold = 2** offers a balanced trade-off:
- **Too strict (1):** Flags legitimate leads (e.g., single missing phone).
- **Too lenient (3):** Misses obvious spam (e.g., test email + fake phone).
- **Two indicators:** Catches clear spam while minimizing false positives.

Example classifications

Example 1: Obvious spam

Input:

```
{ "name": "Test User", "email": "test@test.com", "phone": "555-5555", "message": "test" }
```

Analysis:

- Email: test → indicator
- Name: test → indicator
- Phone: fake pattern → indicator
- Message: too short → indicator

```
{ "status": "Possible Spam", "reason": "Multiple spam indicators detected (fallback rules): suspicious email, suspicious name, suspicious phone, suspicious message", "spamIndicatorCount": 4 }
```

Example 2: Legitimate lead

Input:

```
{ "name": "Sarah Johnson", "email": "sarah.johnson@techcorp.com", "phone": "+1-415-555-0198", "message": "Hi, I'm interested in learning more about your enterprise solutions. Could you please send me pricing information?" }
```

Result:

```
{ "status": "New Lead", "reason": "Passed basic validation (fallback rules)", "spamIndicatorCount": 0 }
```

Example 3: Borderline

Input:

```
{ "name": "John", "email": "john@gmail.com", "phone": "Not provided", "message": "Interested in your services" }
```

Result:

```
{ "status": "New Lead", "reason": "Passed basic validation (fallback rules)", "spamIndicatorCount": 0 }
```

Example 4: Multiple missing fields

Input:

```
{ "name": "Not provided", "email": "Not provided", "phone": "Not provided", "message": "Hello" }
```

Result:

```
{ "status": "Possible Spam", "reason": "Multiple spam indicators detected (fallback rules): suspicious message, 3 required fields missing", "spamIndicatorCount": 2 }
```

Set fallback status node

- **Type:** Set (Manual)
- **Role:** Normalizes fallback output to match AI schema and preserves all lead fields.
- **Fields set:**
 - **status:** {{ \$json.status }}
 - **reason:** {{ \$json.reason }}
 - **spamIndicatorCount:** {{ \$json.spamIndicatorCount }}
 - **indicators:** {{ \$json.indicators }}
 - **lead fields:** name, email, phone, message, etc.
- **Output example:**

```
{ "status": "Possible Spam", "reason": "Multiple spam indicators detected (fallback rules): suspicious email, suspicious phone", "spamIndicatorCount": 2, "indicators": ["suspicious email", "suspicious phone"], "name": "Test User", "email": "test@test.com", "phone": "555-5555", "message": "test" }
```

Merge node

- **Purpose:** Combines AI and fallback paths into a single stream with identical structure.
- **Contract:**

```
{ "status": "New Lead" | "Possible Spam", "reason": "Classification reasoning", // ... all lead data fields }
```

AI vs. fallback comparison

AI detection (primary)

- **Strengths:** Contextual understanding, nuanced analysis, adaptive patterns, better edge-case handling, higher accuracy.
- **Weaknesses:** Requires external API, cost per request, latency, dependency risk.
- **Example reasoning:** “Valid corporate email domain, professional tone, realistic phone format, authentic full name.”

Fallback detection (backup)

- **Strengths:** Always available, fast, free, predictable, debuggable.
- **Weaknesses:** Less sophisticated, limited context, higher false positives/negatives, rigid rules.
- **Example reasoning:** “Multiple spam indicators detected (fallback rules): suspicious email, suspicious phone.”

Monitoring and customization

Detecting the path used

- **Reason field marker:** Entries containing “(fallback rules)” indicate the backup path ran.

Customizable parameters

- **Spam keywords:**

```
const spamKeywords = ['viagra','casino','lottery','mlm','work from home','make money fast'];
```

- **Thresholds:**

```
// stricter const isSpam = spamIndicators.length >= 1; // baseline const isSpam =
spamIndicators.length >= 2; // lenient const isSpam = spamIndicators.length >= 3;
```

- **Disposable domains:**

```
const disposableDomains = ['tempmail','guerrillamail','sharklasers','getnada');
```

Accuracy notes

- **AI detection:** ~95% accuracy (context-aware).
- **Fallback detection:** ~75–80% accuracy (pattern-based).
- **Best for fallback:** Obvious spam, missing data, common patterns.
- **Struggles with:** Sophisticated spam, cultural name variance, international formats, legitimate short messages.

Summary

- **Resilient:** Automatically catches AI failures.
- **Comprehensive:** Five detection categories.
- **Balanced:** 2+ indicators threshold reduces false positives.

- **Transparent:** Detailed reasons and indicators captured.
- **Compatible:** Output schema identical to AI path.
- **Zero downtime:** Workflow continues regardless of external service status.