

Leakage Detection on Infrared Images of Under Pipes

hiher

UDADA

TUJ

I an

Mihrimah GÖKNAR Supervisor: Prof. Dr. Reda Alhajj

INTRODUCTION

Leakage in pipelines carrying water or hydrocarbon fluids is one of the risks that can result in serious injuries, environmental disasters, and a significant economic effect. A preventive, non-destructive inspection method can help to avoid a situation like this.

RESULTS

	Age (AG)	MaterialType (MT)	Size (SZ)	Installation Quality (IQ)	SurfaceType (SR)	Ground WaterLe vel (GW)	SoilType (SL)	C-Factor (CF)	BreakRate (BR)	WaterQuality (WQ)	Performance Index (PI)	FailureProb (FP)	Failure (F)
pipe1	43	Concrete	184	Good	Asphalt	Moderate	Non- aggressive	106	0.340569	Fair	6.6031	0.288634	N
pipe2	11	Concrete	370	Good	Unpaved	Deep	Non- aggressive	58	0.036081	Good	8.83205	0.035437	N
pipe3	64	PVC	123	Poor	Asphalt	Deep	Aggressive	159	0.05535	Fair	6.01225	0.053846	N
pipe4	53	Asbestos	420	Poor	Unpaved	Deep	Moderate	196	0.556447	Fair	6.0593	0.426758	N
	17	A .1	402	Γ.'.		Madanata		10	0.25507	Carl	5 (120	0.000076	N

1 ooi pain Moderaic Aggressive

OBJECTIVE

(thermal Infrared technology imaging technology) is a non-destructive assessment for detecting and assessing the severity of a leak employing sensors to identify the by wavelength of infrared light emitted from an object's surface.



Figure 8: Example Thermal Image data. a-b is No Leak, c-d Leak cases





0.3330

UUUU

J.01J0

0.4/00/



METHODOLOGY



Figure 11: Accuracy and Loss Plot for 100 epochs

A hybrid model is developed to identify leakages in 2 phase to increase the performance of the system. In the first phase, the metadata of the pipes is being tested on the decision tree and then if the pipe is classified as 'Failure' then the thermal image of that pipe is acquired and sent to the CNN model to detect the leakage.

Meta Data

x(2)

Age(AG)	106
Size(SZ)	458
C-Factor(CF)	86
BreakRate(BR)	0.332498
PerformanceIndex(PI)	4.67285
FailureProb(FP)	0.28287
MaterialType(MT)	3
InstallationQuality(IQ)	2
SurfaceType(SR)	Ø
GroundWaterLevel(GW)	2
SoilType(SL)	Ø
WaterQuality(WQ)	1
Image ID	image4213
and the second	

Age(AG)	52	
Size(SZ)	235	
C-Factor(CF)	49	
BreakRate(BR)	0.242485	
PerformanceIndex(PI)	4.72095	
FailureProb(FP)	0.215325	
MaterialType(MT)	4	
InstallationQuality(IQ)	2	
SurfaceType(SR)	1	
GroundWaterLevel(GW)	2	
SoilType(SL)	Ø	
WaterQuality(WQ)	Θ	
Image ID	image2097	

CONCLUSION

If this project could be adapted to real world water, gas and other hydrocarbon fluid networks, it will not only contribute economically (reducing the cost for digging and changing, repairing the pipes), to several institutions that operate and control buried pipes, but also ensure that leaks in pipes are detected early, without the need for open inspection and before they become harmful to the environment and humanity.



Figure 13: Sample output of leaking and not leaking case

1. Manekiya, M. H., Arulmozhivarman, P. (2016). Leakage detection and estimation using IR thermography, 2016 International Conference on Communication and Signal Processing (ICCSP), 1516-1519. 2. Alvarado, E. P. C., Cabrera, D. A., García, R. P., Sebastián, J. I. (2014). Identification of Buried Pipes Using Thermal Images and Data Mining. Procedia Engineering, 89, 1445–1451. 3. Chanati, H. E., El-Abbasy, M. S., Mosleh, F., Al-Derham, H. (2015). Multi-Criteria Decision Making Models for Water Pipelines. American Society of Civil Engineers, 30,4. 4. S. A. Thomas, (2021). Combining Image Features and Patient Metadata to Enhance Transfer Learning, 43rd Annual International Conference of the IEEE Engineering in Medicine & Biology Society (EMBC), Mexico, 2660-2663.