### FULLY-NEURAL APPROACH TO HEAVY VEHICLE DETECTION ON BRIDGES USING A SINGLE STRAIN SENSOR

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# **INTRODUCTION**

#### ACCUMULATED VEHICLE LOADS ON BRIDGES

- may destroy gradually more than 700,000 road bridges in Japan.
- must be collected automatically for structural health monitoring.





#### PAVEMENT WEIGH-IN-MOTION (P-WIM)

- is installed on the pavement surface somewhere in the road network.
- estimates axle loads and collect evidence without stopping vehicles.





#### BRIDGE WEIGH-IN-MOTION (B-WIM)

- exploits bridge components, e.g., main girders, as weighing scales.
- Peak values of the strain responses contain axle load information:





#### **COMPLEMENTARITY: P-WIM AND B-WIM**

- to capture overloaded vehicles making a detour to avoid P-WIMs,
- so that no vehicles running in the road network can break the law:







## **CONVENTIONAL B-WIM**

#### **B-WIM: KERNEL-FITTING APPROACH**

- estimates axle loads by fitting a unit response to the strain signal:
- The unit response is called 'influence line', unique to each bridge.





#### **B-WIM: AREA-BASED APPROACH**

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- kernel fitting requires axle positions; sometimes difficult to obtain.
- Gross weight W is calculatable from product of area A and speed v:

$$s(t) = \sum_{n=1}^{N} w_n i(vt - l_n), \qquad s(t): \text{raw strain sequence in time domain,} \\ i(x): \text{ influence line, N: number of axles,} \\ l_n: n-\text{th axle position, } w_n: n-\text{th axle load.} \end{cases}$$
$$A = \int_{-\infty}^{\infty} s(t) dt = \sum_{n=1}^{N} w_n \int_{-\infty}^{\infty} i(vt - l_n) dt. = \frac{W}{v} \int_{-\infty}^{\infty} i(x - l_n) dx.$$

Area-based approach is easier and widely applied to many bridges.



### DIFFICULTY OF STRAIN MEASUREMENT

- B-WIM utilizes multiple strain sensors for accurate load estimation:
- It takes time and effort to install many strain sensors on the bridge.





### PARAMETER SENSITIVITY OF INFLUENCE LINE

- may degrade the accuracy of the axle-load estimation via B-WIM.
- B-WIM must consider accurate vehicle movement on the bridge:







## OPREVIOUS WORK IN 2019

#### PREVIOUS WORK: DEEP-SENSING APPROACH

- was proposed as a single-sensor vehicle detector for B-WIM in 2019.
- detected vehicles and their properties using a single strain sensor.





### **FULLY-NEURAL B-WIM**

### **CONCEPT: TRAINING B-WIMS USING P-WIMS**

- P-WIMs provide B-WIMs with vehicle IDs with known axle loads.
- B-WIMs learn influence lines by using the axle load information.



#### **CONCEPT: TRAINING B-WIMS USING P-WIMS**

#### P-WIMs

- detect heavy vehicles,
- extract features from video data,
- for vehicle reidentification (Re-ID).

#### **B-WIMs**

- retrieve load data by vehicle Re-ID,
- learn responses to known vehicles,
- predict loads for unknown vehicles.



#### **PROPOSAL: FULLY-NEURAL B-WIM**

• CNN as a vehicle detector and load estimator in a multi-task fashion:





#### NEURAL NETWORK ARCHITECTURE

Residual CNN; 1 convolution, 8 plain residual blocks, 3 linear layers:





#### **6 PREDICTION TASKS**







## **EXPERIMENTAL RESULTS**

#### EXPERIMENTAL SETUP

- 11 strain sensors, 2 cameras on a steel bridge in an expressway:
- We retrieved 5,923 heavy vehicles with known axle loads by Re-ID.



- Fully-neural BWIM used a single sensor installed beneath the deck.
- Area-based BWIM used all sensors at decks, girders, and v-stiffeners.



#### **EXPERIMENTAL SETUP: AREA-BASED B-WIM**

• Our special implementation estimated axle loads via 3 linear layers:





#### FULLY-NEURAL B-WIM VS AREA-BASED B-WIM





#### TASK5 VS CAMERA-ASSISTED TASK6





#### SUMMARY

Additional features from cameras

• were effective to improve axle-load accuracy.

vs Area-based BWIM

• equivalent although proposal utilized only a single sensor.

Future work

• combine the 11 sensors on decks, girders, and v-stiffeners.





### **OTHANK YOU!**