Development of a low frequency vibration isolation system for KAGRA, and study of the localization of coalescing binaries with a hierarchical network of gravitational wave detectors.

Master's thesis defense 35-156218 **Yoshinori Fujii**

Contents

75°

45°

30°

15°

0°

22h 20 18 16h 14h 12h 10h 8h 6h 4h 2h

-30°

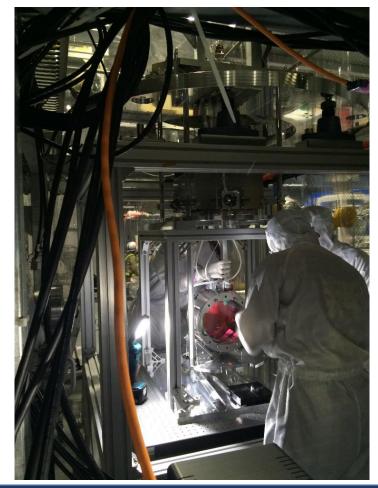
-45°

-60°

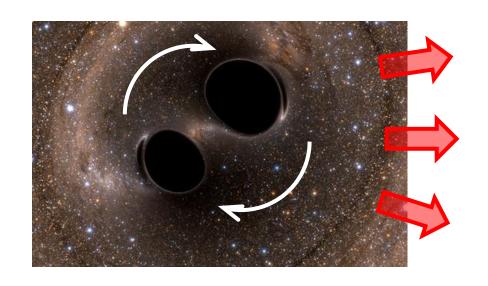
1. Source localization

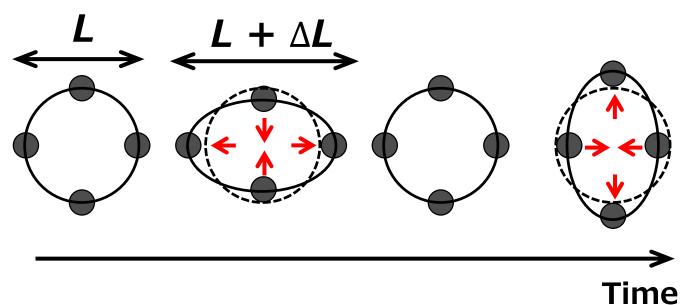


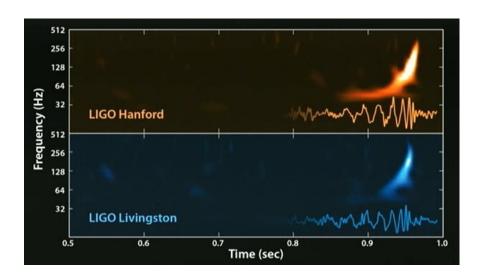
2. Detector development



Gravitational wave







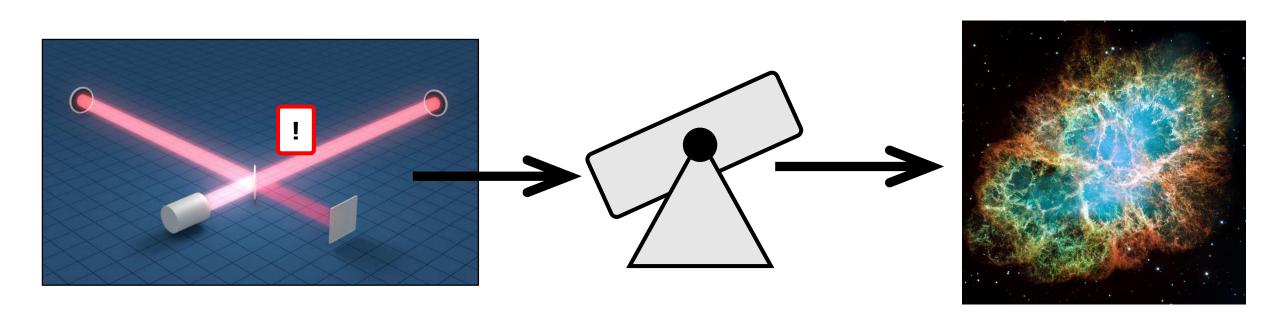
First detection! done!

→ New astronomy!

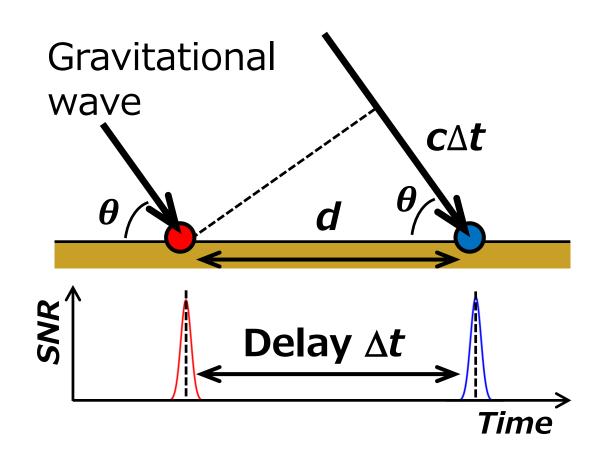
For starting astronomy,

Source localization.

for follow-up observation.



From where?



Time delay

Localization

$$\Delta t \longrightarrow \theta$$

$$\theta = \cos^{-1}\left(\frac{c\Delta t}{d}\right)$$

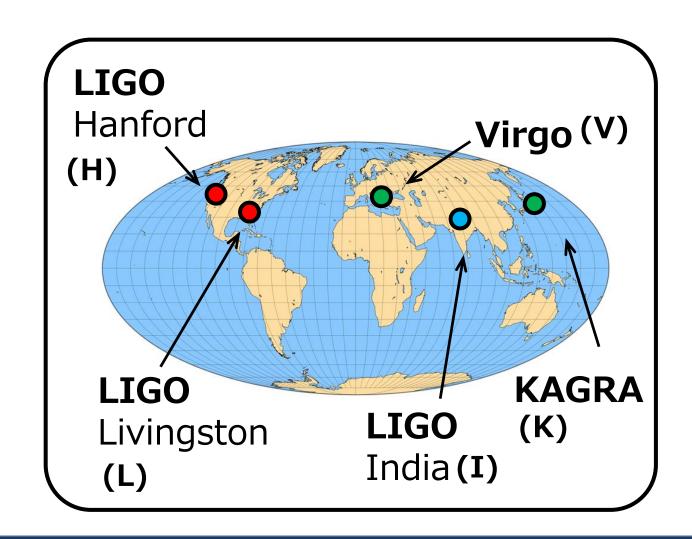
For localization, we want...

→ Several detectors!

Continuous observation

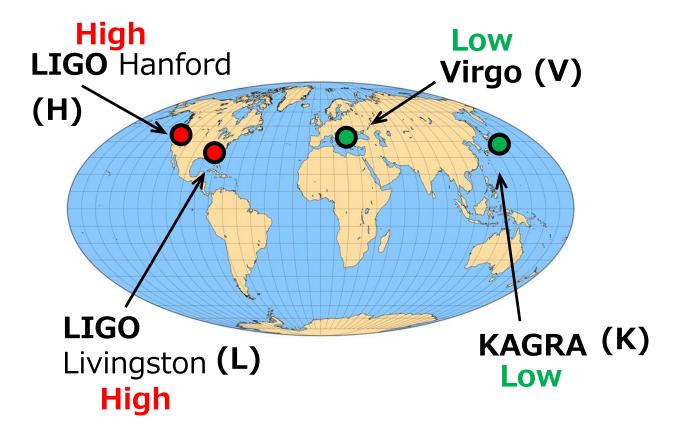
All sky coverage

Precise localization!



Different sensitivities.. OK?

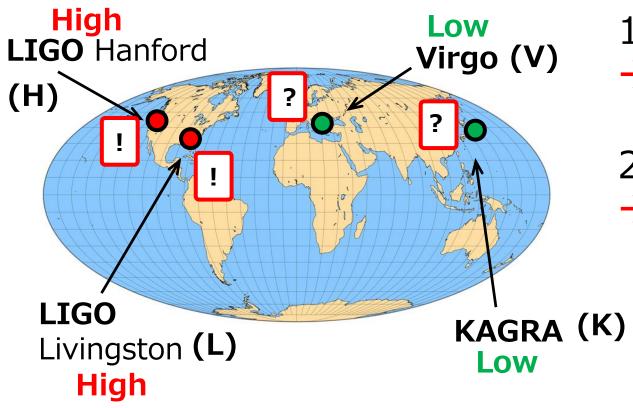
Ex.) SNR $> 5 \rightarrow$ detection



(At the beginning)

Different sensitivities.. OK?

Ex.) SNR $> 5 \rightarrow$ detection

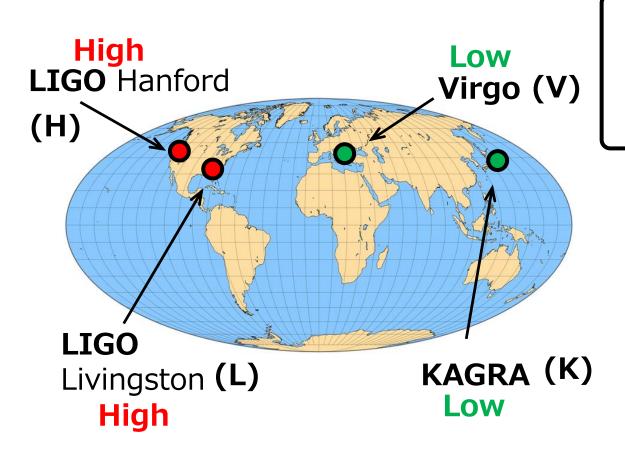


- 1) Triple (or more) coincidence
- → Rare
- 2) Double coincidence
- → Not precise localization

(At the beginning)

Hierarchical network search





1. A trigger is detected by *high sensitivity* detectors with *high SNR threshold*



2. Analyze

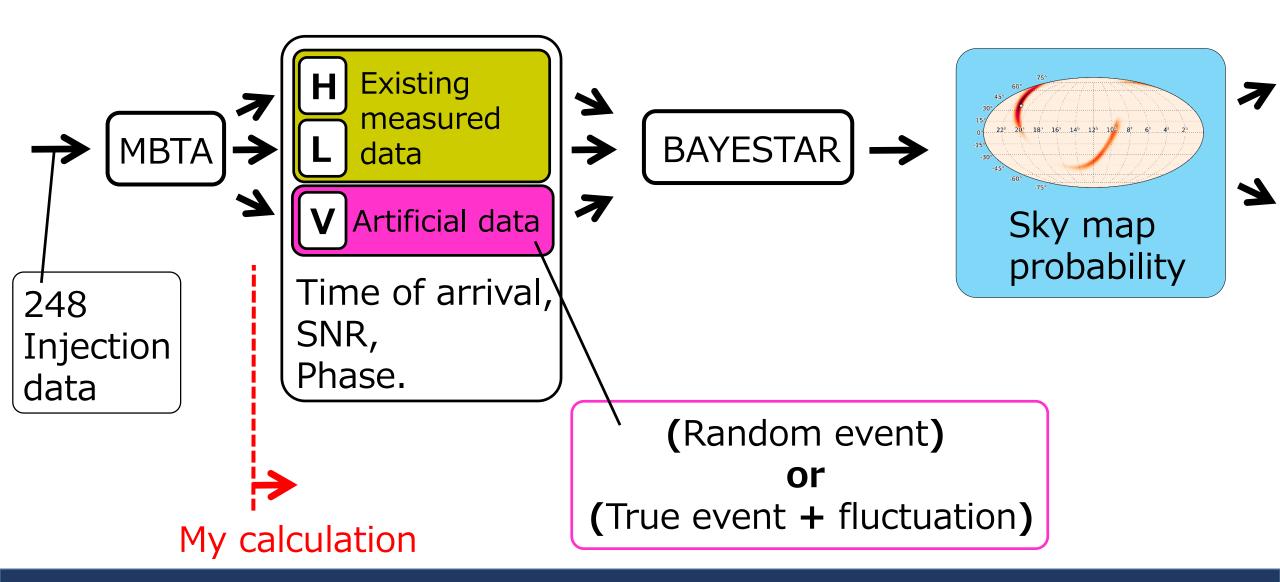
low sensitivity detectors
with low SNR threshold
using small time window.

Assumption in calculation

1. GW-EM pipeline for GWs from CBC 2nd BAYESTAR **MBTA** detectors telescopes Signal Sky map **Event info: Compact** probability SNR, **Binary** arrival time, Coalescence etc.

2. Two LIGOs (70 Mpc), Virgo (20 Mpc) — High sensitivity × 2 / Low sensitivity × 1

Calculation main flow 1

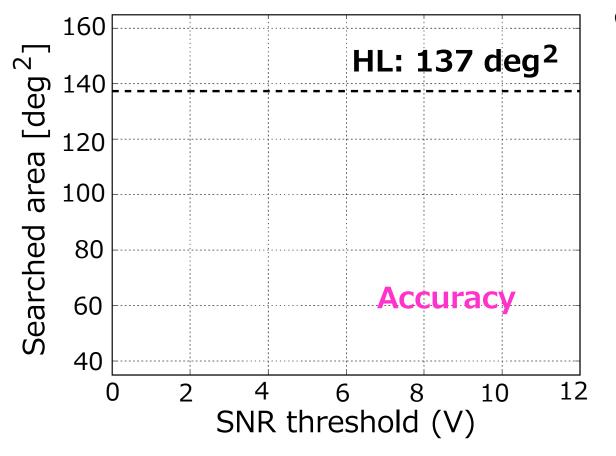


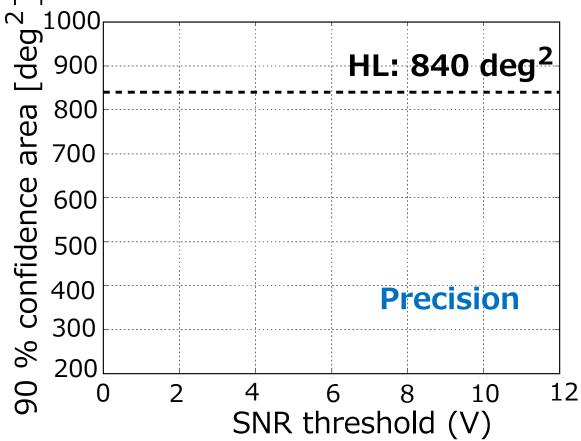
Calculation main flow 2

Localization performance Histograms from 248 1) Accuracy events. → Searched $FAR < 10^{-7} Hz$ area (deg²) median values searched area (deg2) 2) Precision $FAR \le 10^{-7} Hz$ **→** 90 % confidence area (deg²)

Expected performance, HLV

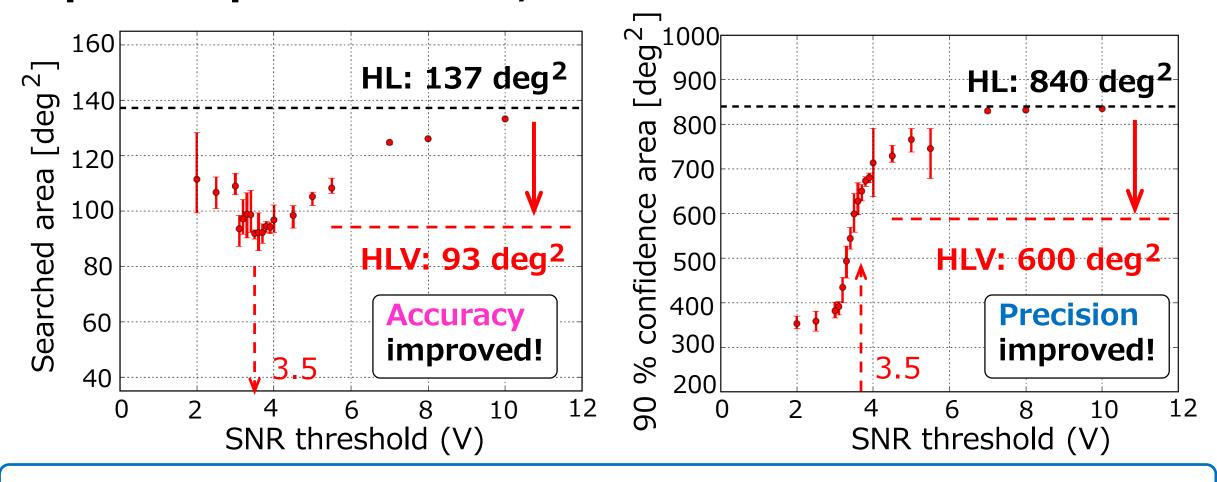
(SNR threshold for H, L = 5.)





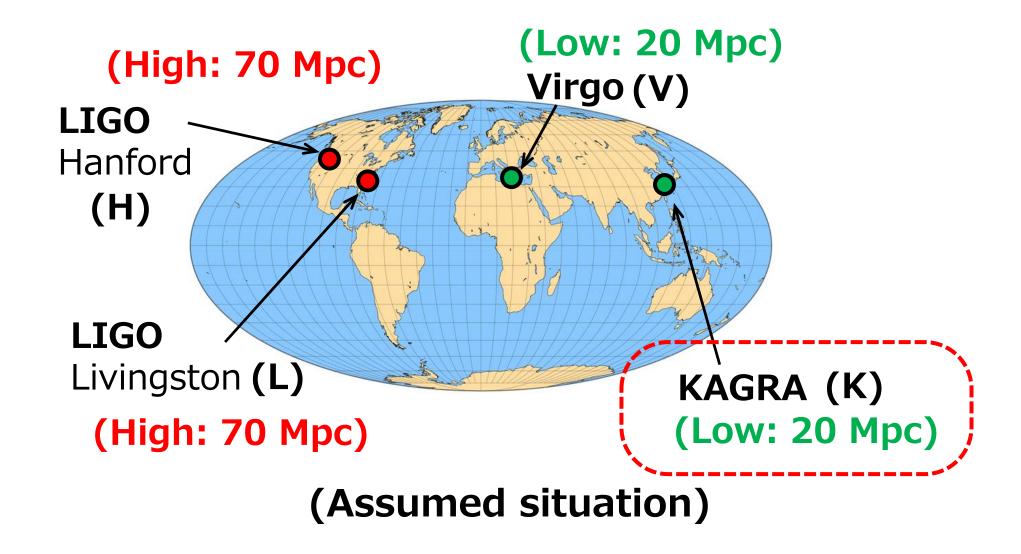
Expected performance, HLV

(SNR threshold for H, L = 5.)



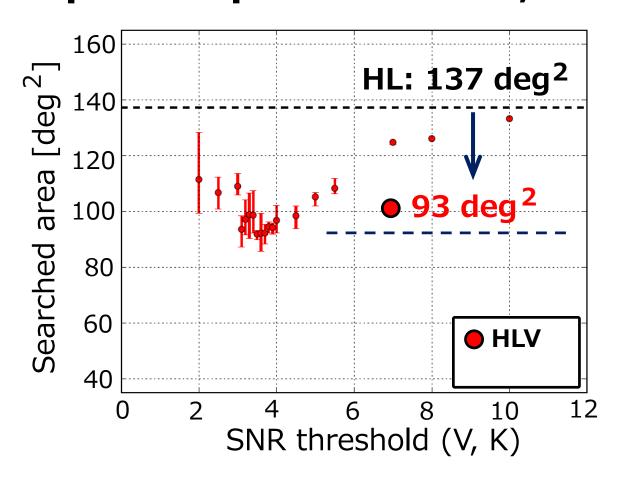
→ By including low sensitivity detector, errors on sky maps will be reduced by a factor of ~ 0.7 than HL.

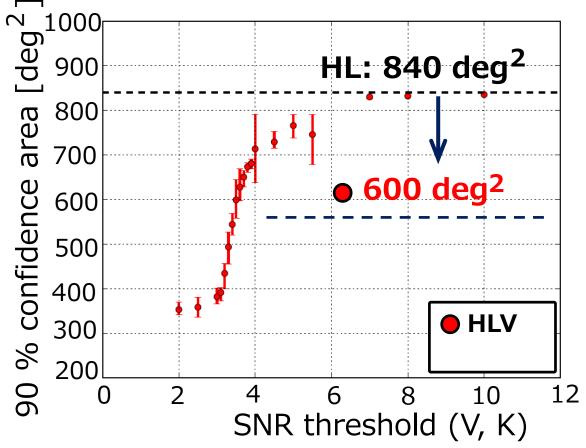
How about 4 detectors, HLVK?



Expected performance, HLVK

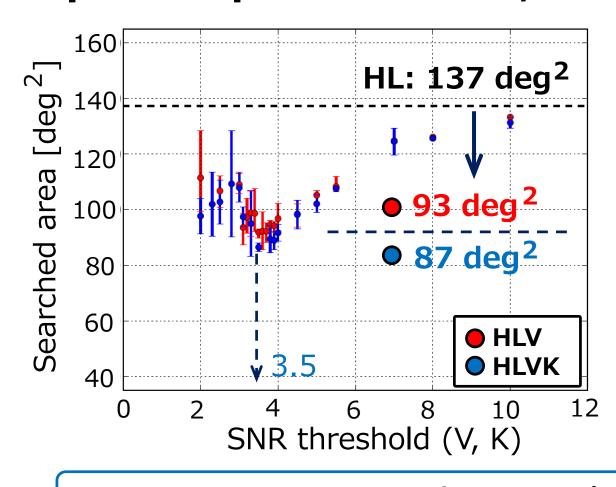
(SNR threshold for H, L = 5.)

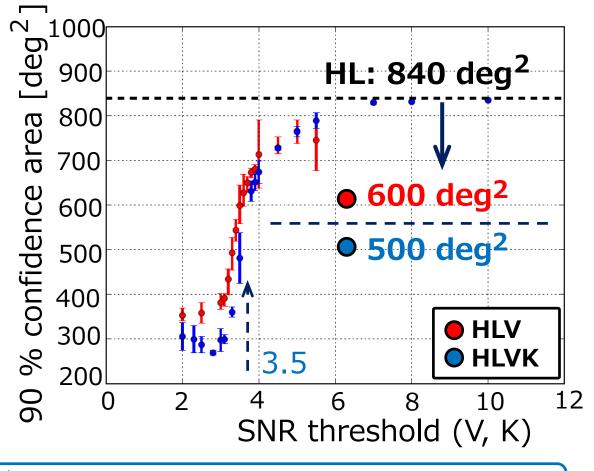




Expected performance, HLVK

(SNR threshold for H, L = 5.)





Accuracy → Not so improved..

Precision → improved!



4th detector contributes to improvement!

Summary 1

A localization with a hierarchical network is demonstrated.

In network by 3 GW detectors (70 Mpc ×2 and 20Mpc),

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Accuracy Precision are reduced by a factor of ~ 0.7 than HL.
```

→ Low sensitivity detector can contribute!

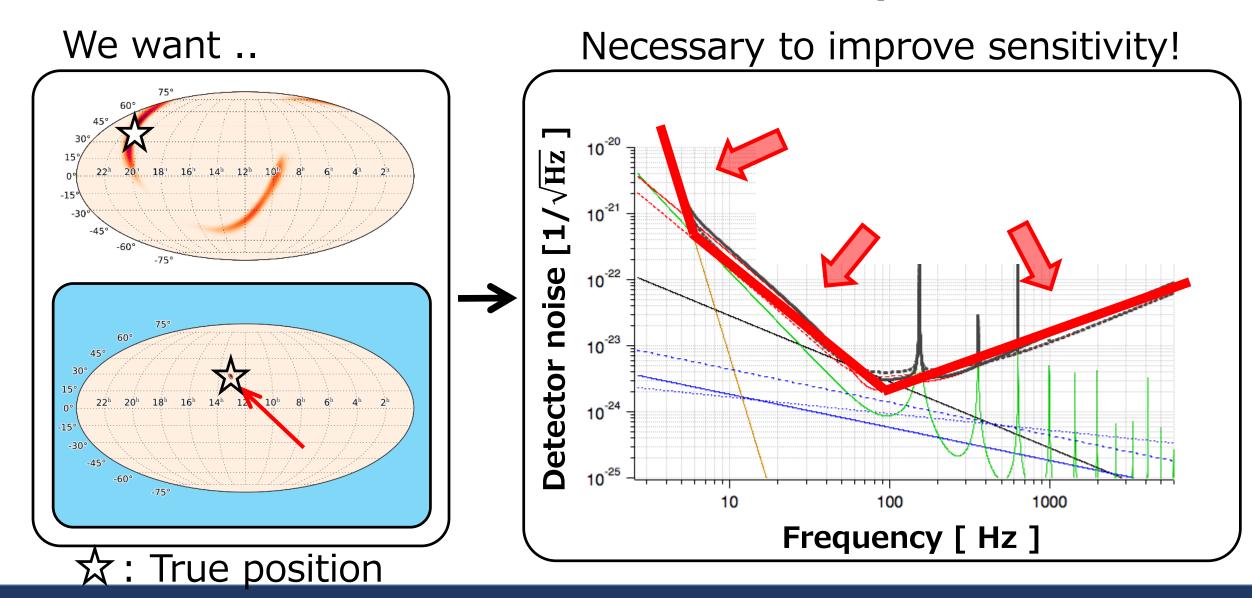
In network by 4 GW detectors (70 Mpc \times 2 and 20Mpc \times 2),

Accuracy: HLV ~ HLVK

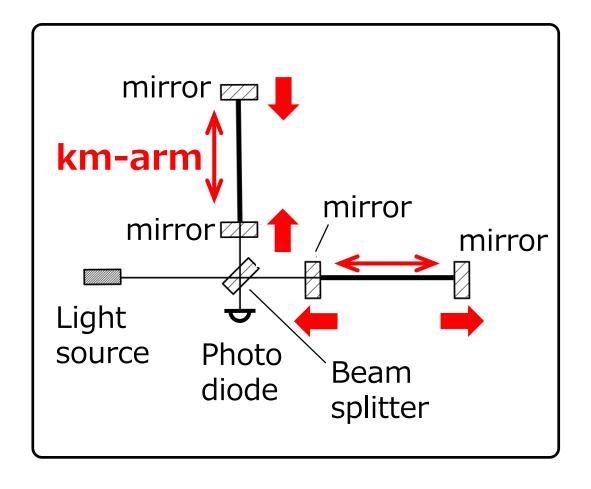
Precision: reduced by a factor of ~ 0.8 than HLV.

- → 4th detector can contributes!
- → useful for follow-up observation!

Source localization → detector development



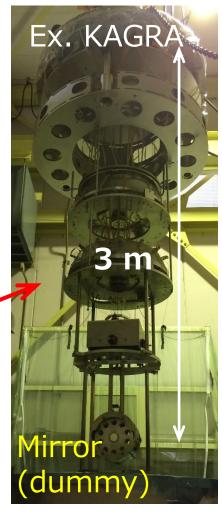
Gravitational wave detector



- 1) Michelson-based interferometer
- 2) Fabry-Perot cavities
- 3) km-arm



4) Suspended core optics



Detector noise

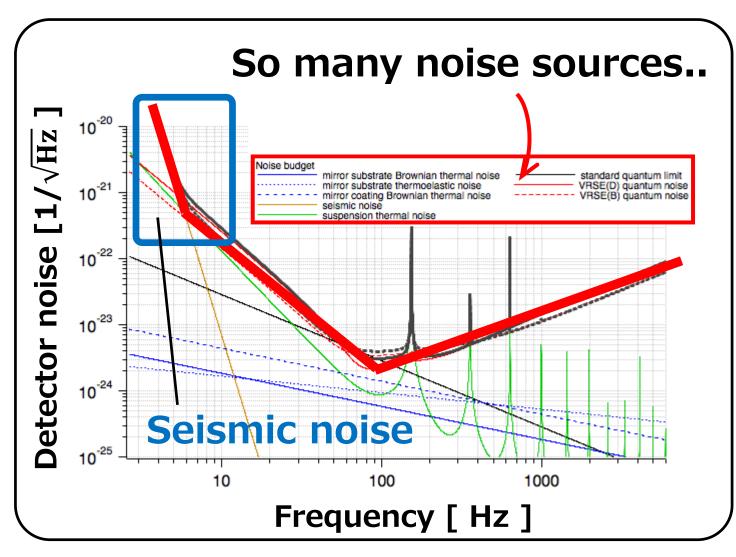
- Quantum noise
- Thermal noise

. . .

- Seismic noise

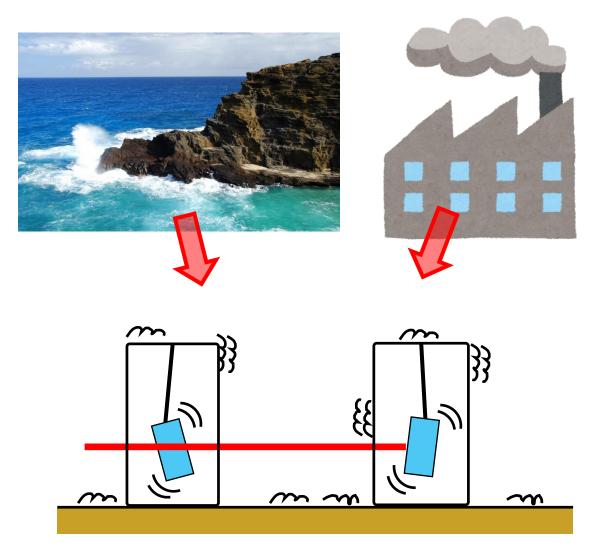
mirror oscillation

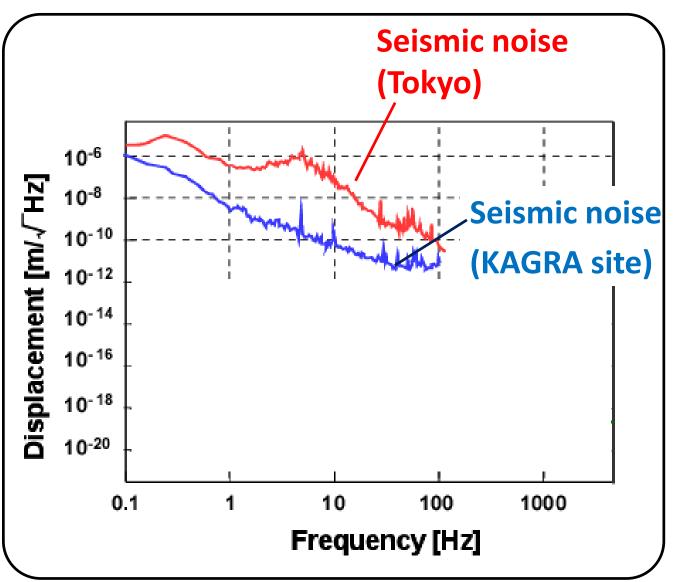
→ Necessary to suppress



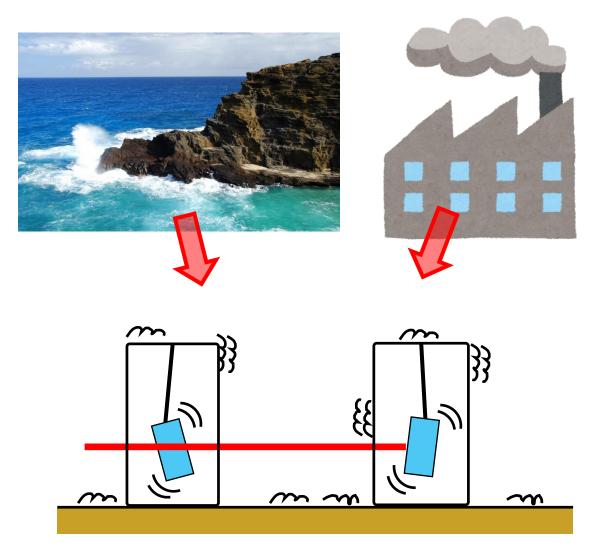
In case of KAGRA

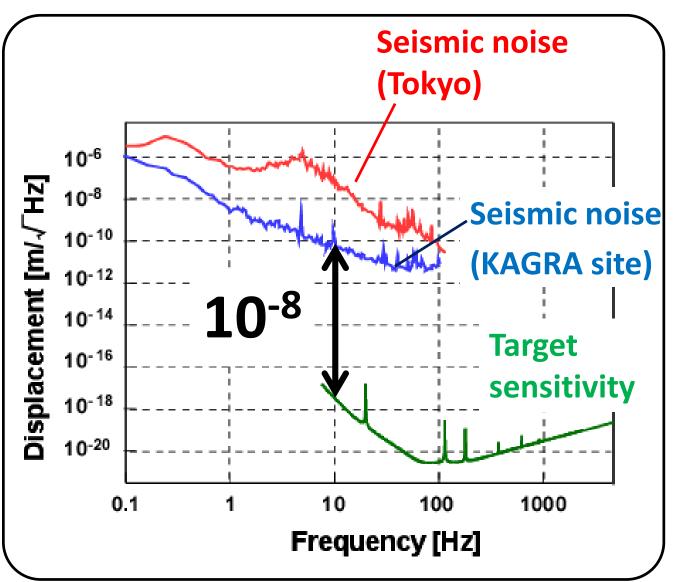
Seismic noise



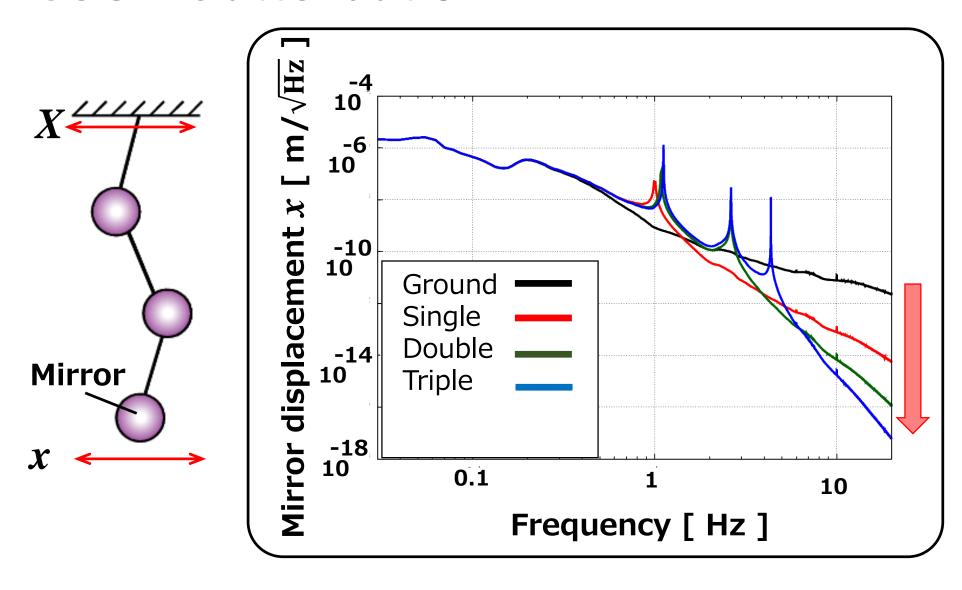


Seismic noise

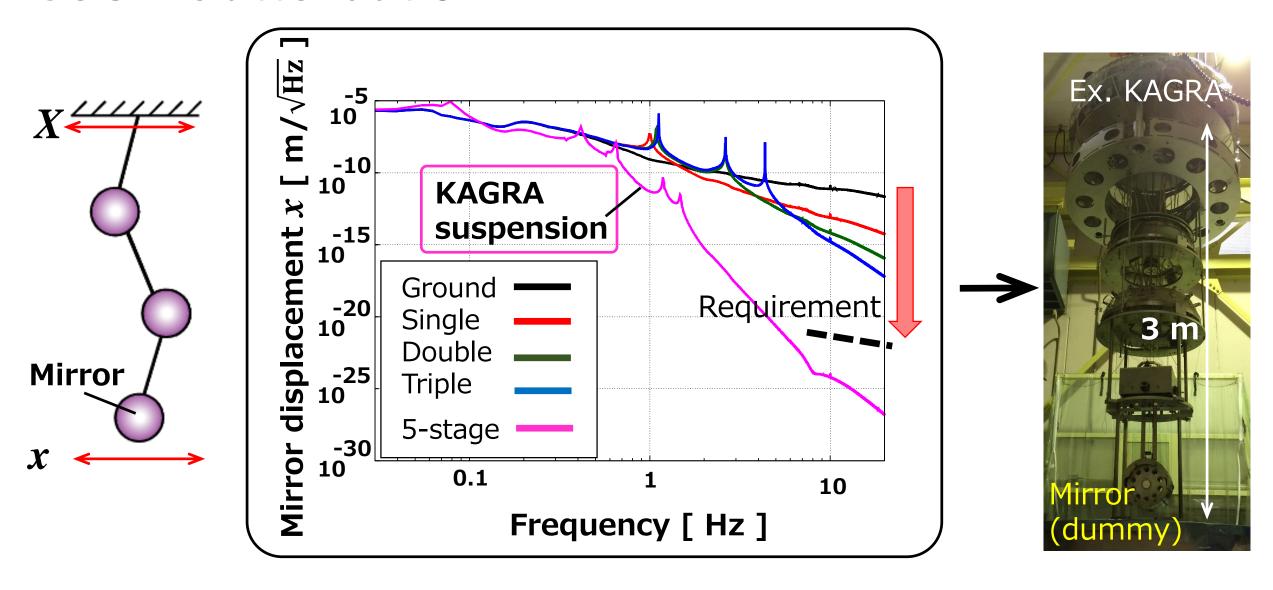




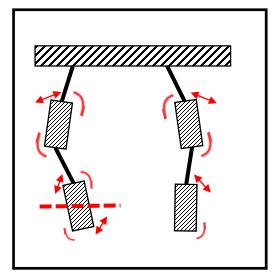
Seismic attenuation



Seismic attenuation



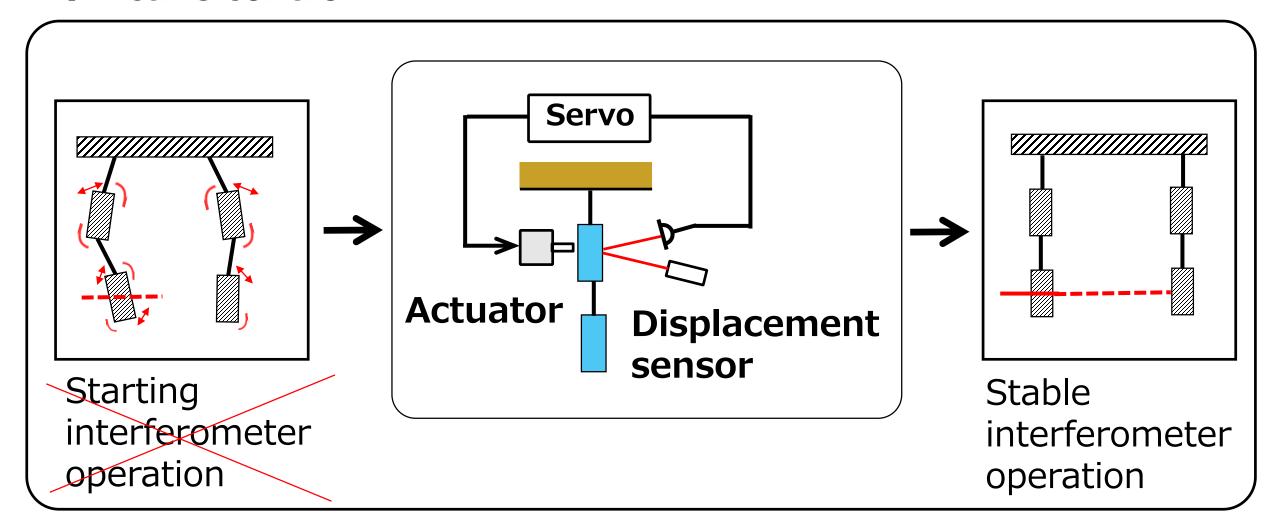
Resonance damping



Starting interferometer operation

Resonance damping

→ Active control

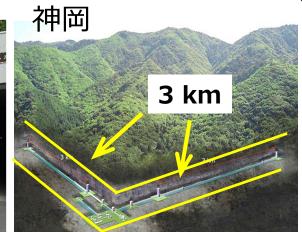


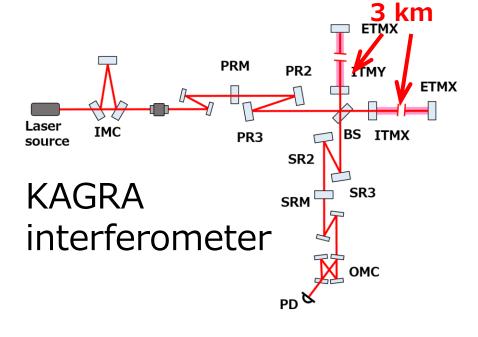
KAGRA project

KAGRA detector

- 1) Japanese detector
- 2) now being developed
- 3) underground







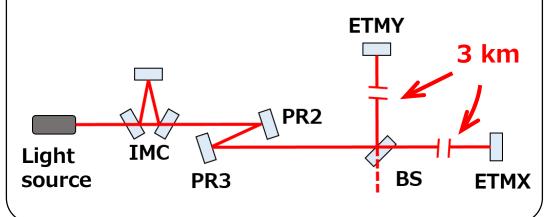
KAGRA project

KAGRA detector

- 1) Japanese detector
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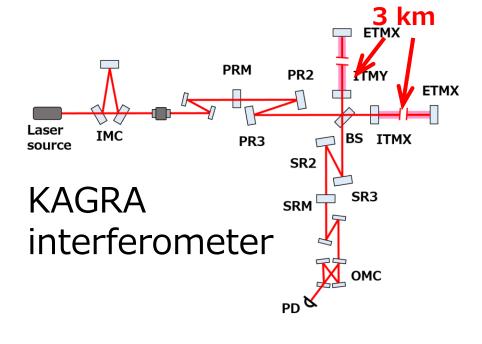


- 1) test run in 2016
- 2) Simple interferometer









iKAGRA suspension development

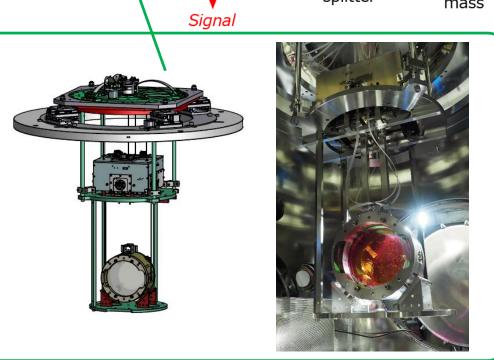
Development work:

- 1) Assembly
- 2) Measurement
- 3) Upgrade for final phase

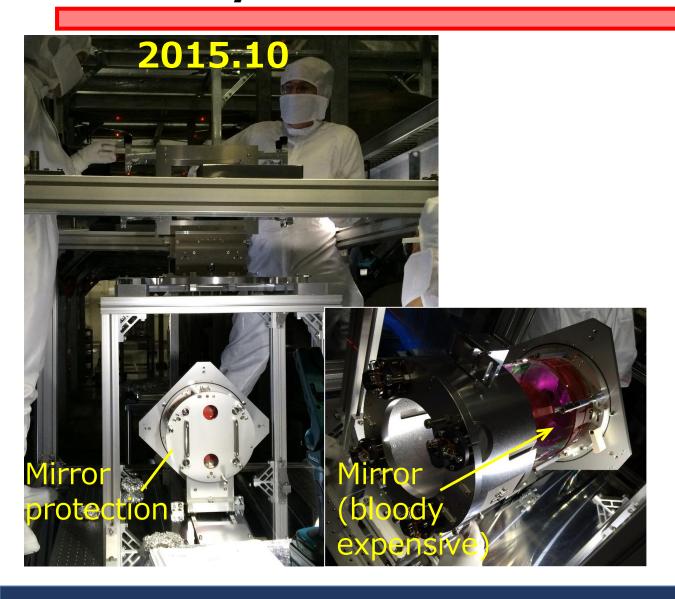
Input mode cleaner PR2 Beam splitter Signal Find test mass Beam splitter Signal

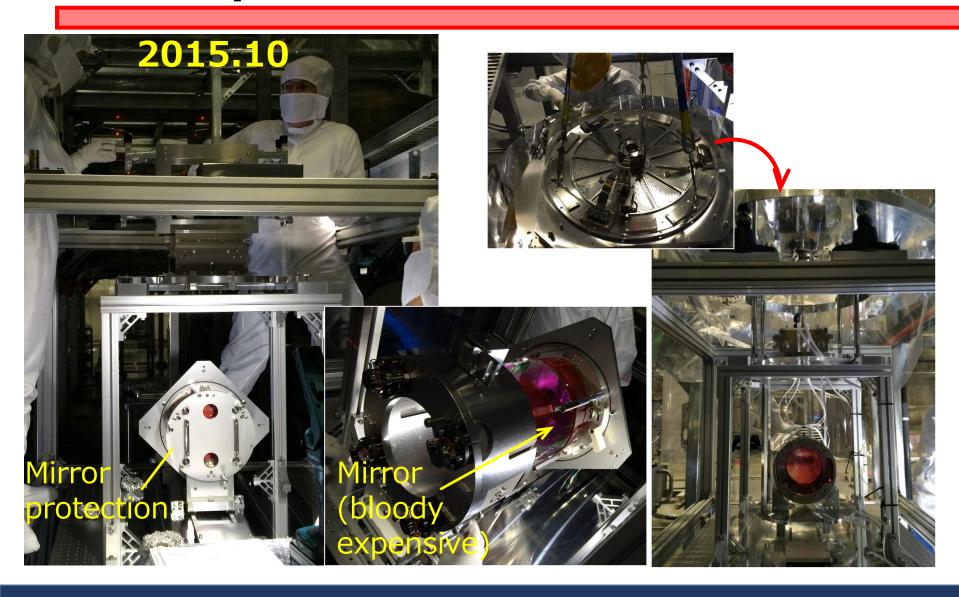
iKAGRA suspension:

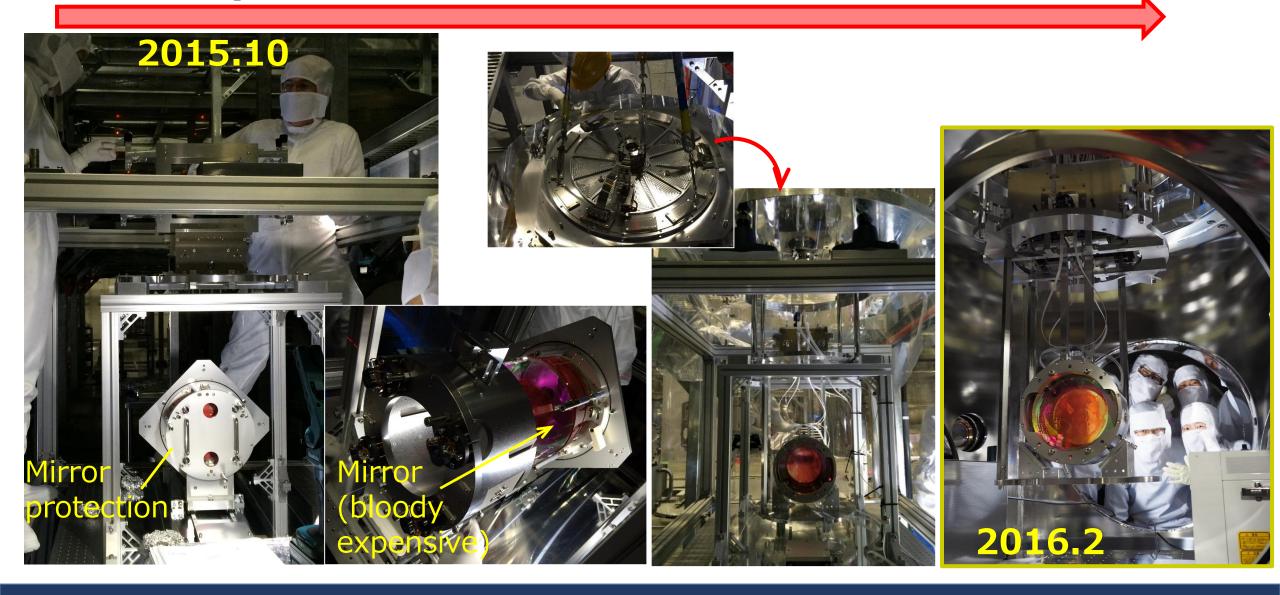
Alignment mirror of iKAGRA for initial alignment for stable operation.



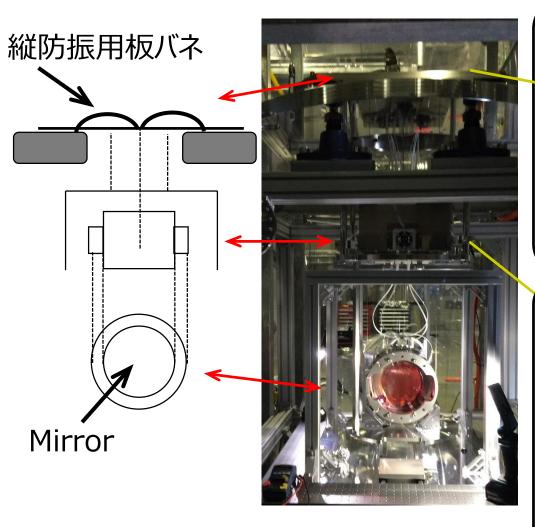








Sensors and actuators

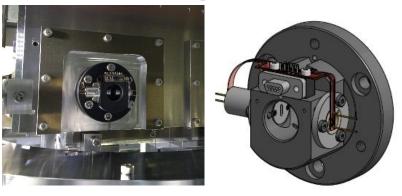


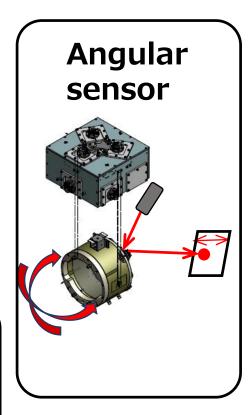
iKAGRA suspension

Displacement sensor and coil-magnet actuator 1

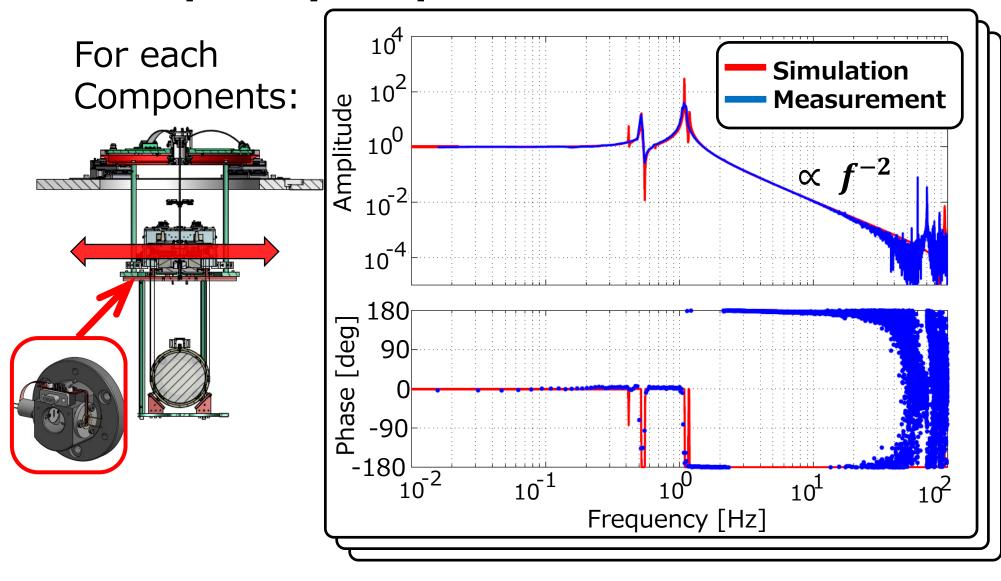


Displacement sensor and coil-magnet actuator 2



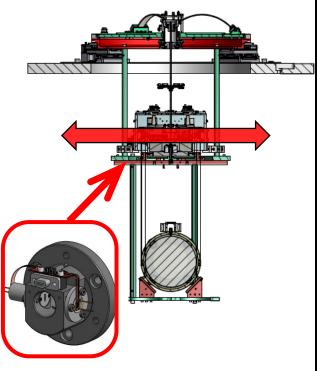


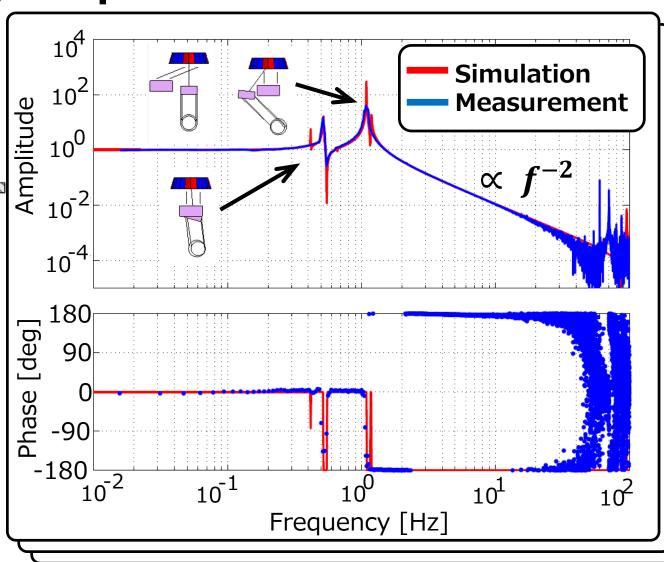
1. Frequency response



1. Frequency response

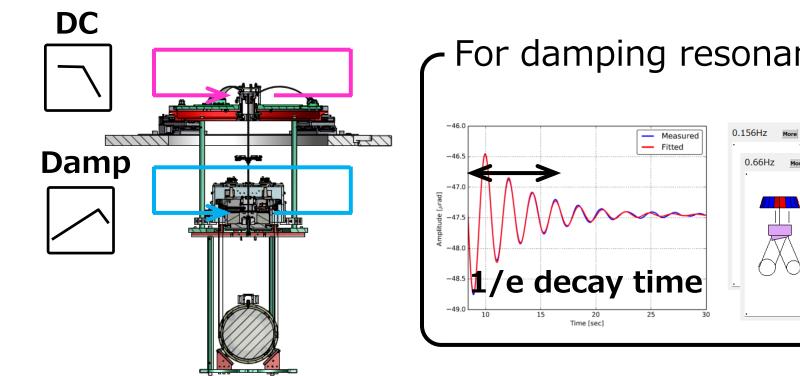
For each Components:

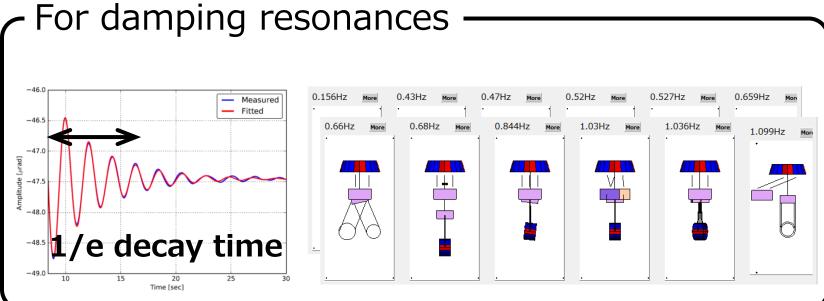




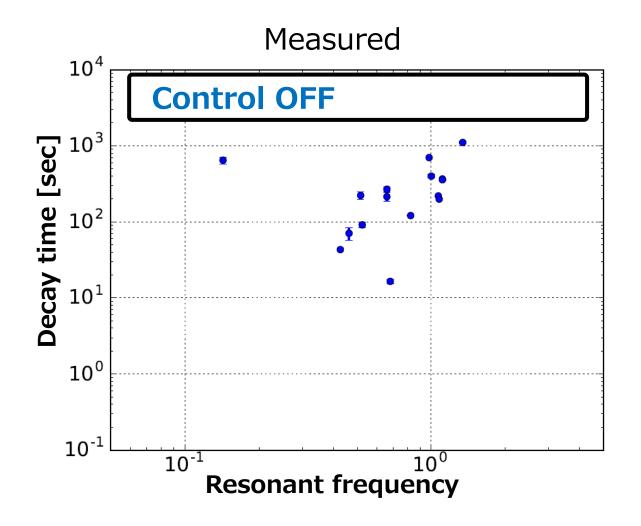
- 1)Measurement
- → Consistent with Simulation
- 2) Components
- → Pendulum

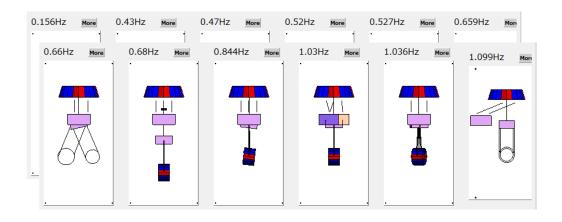
2. Decay time

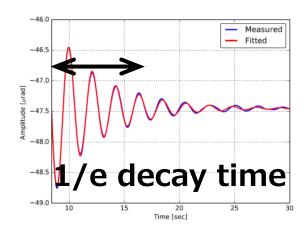




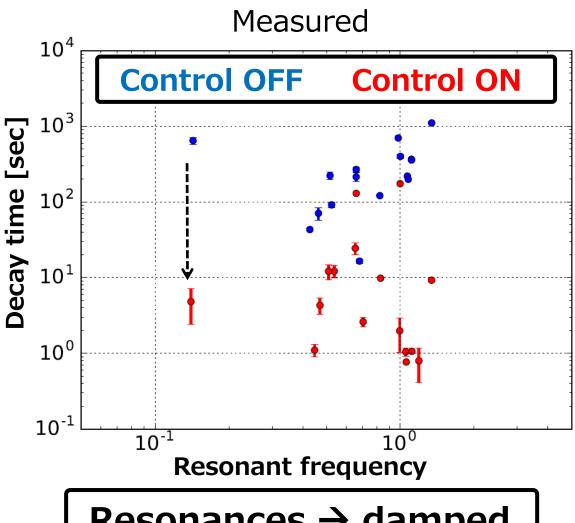
2. Decay time without damping

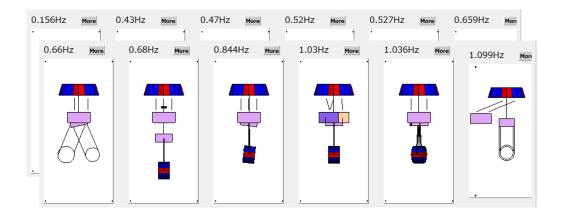


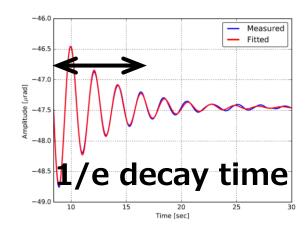




2. Decay time with damping

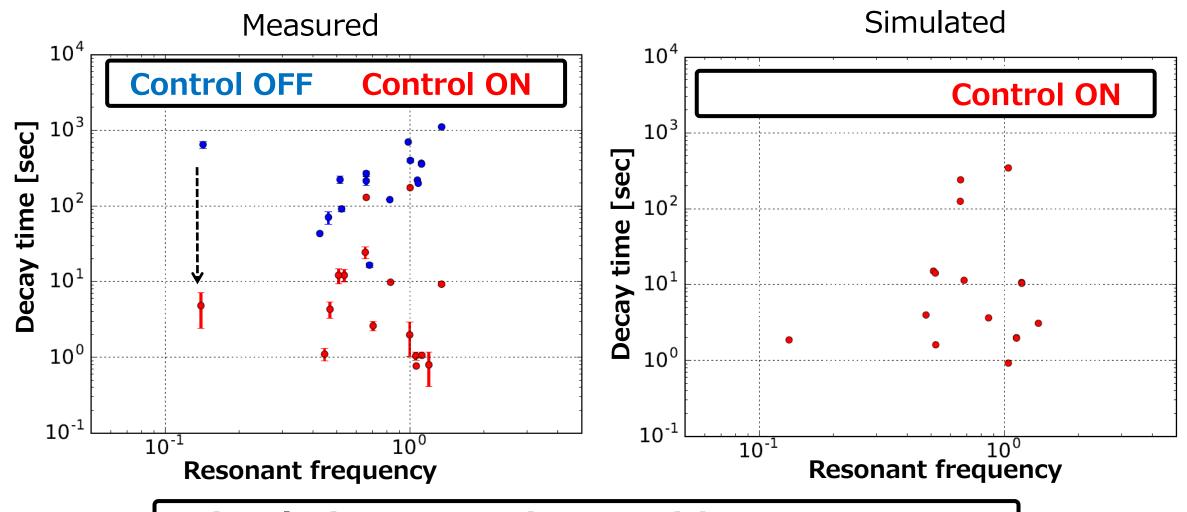






Resonances → damped

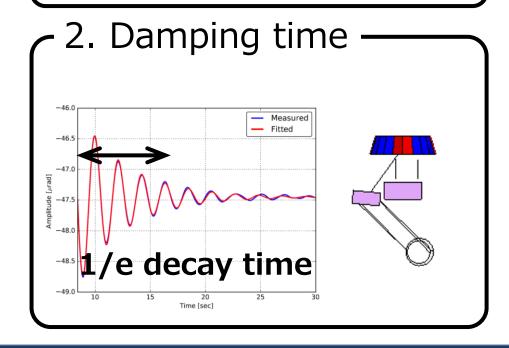
2. Decay time with damping



Simulation → consistent with measurement

Measurement:

C 1. Frequency response -



Suspension

→ Pendulum

Resonances

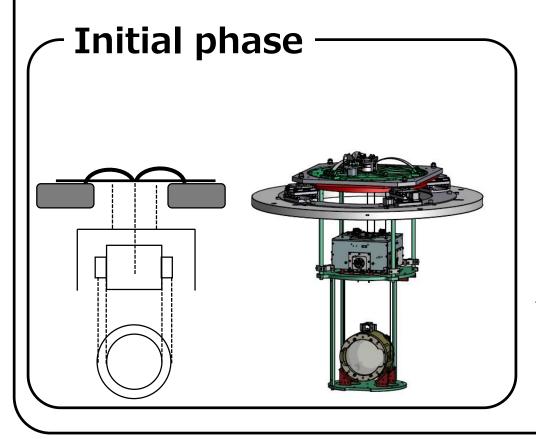
→ Damped

Measurement

→ Consistent with simulation

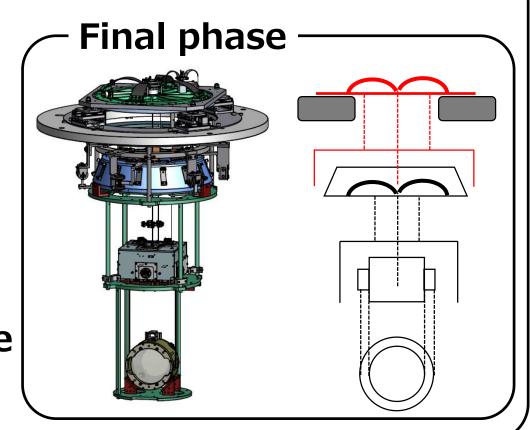
Upgrade: iKAGRA → final KAGRA

In order to meet final requirements:



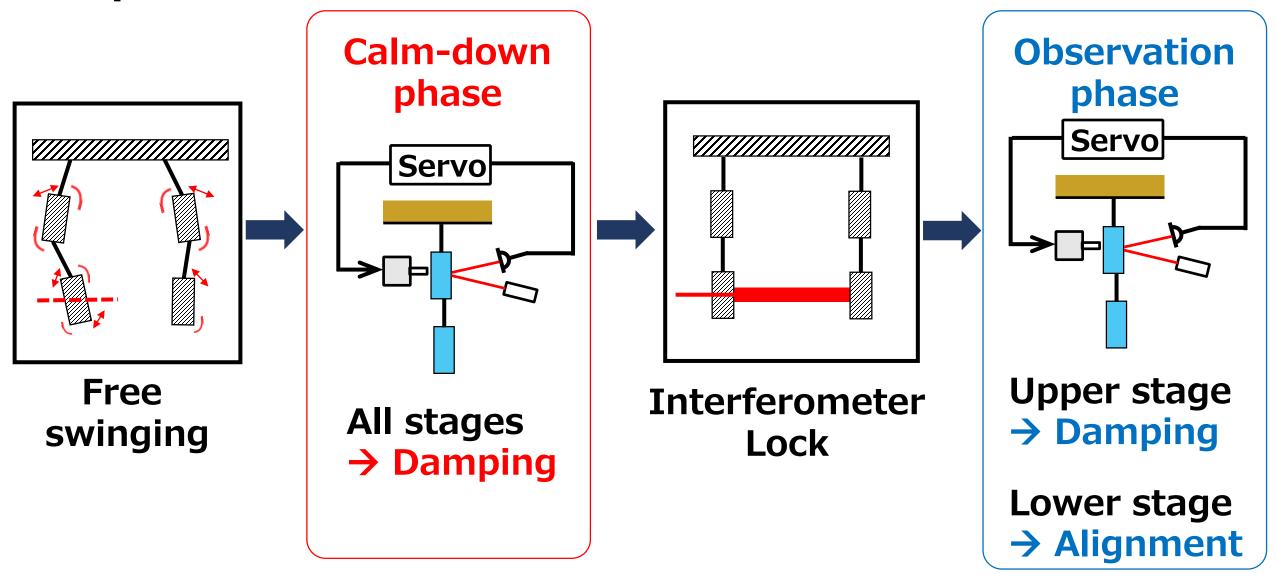


Add one more stage



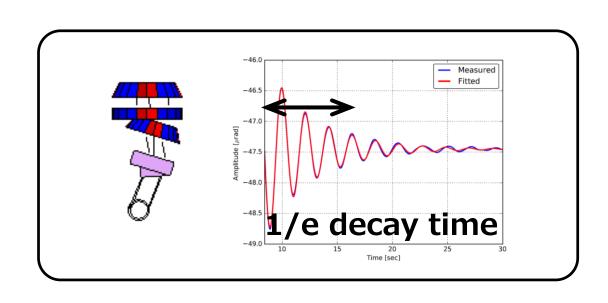
→ Design active control systems.

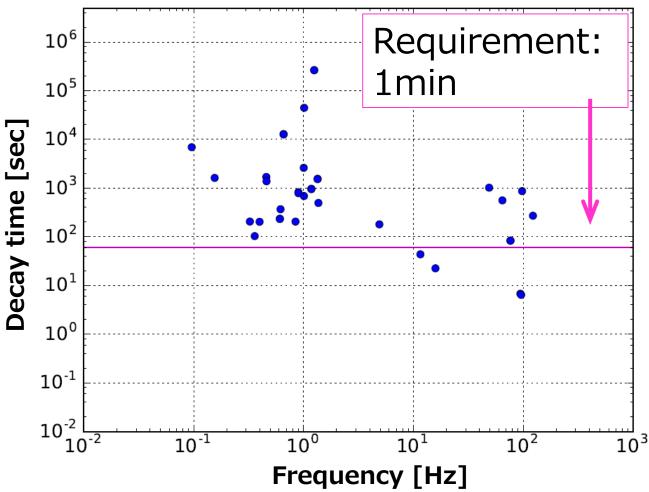
Steps for observation



Calm-down phase:
Suppress large disturbance

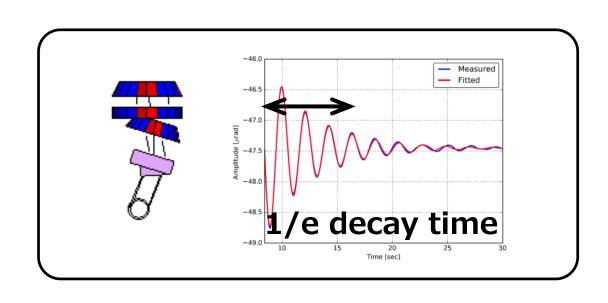
Control OFF

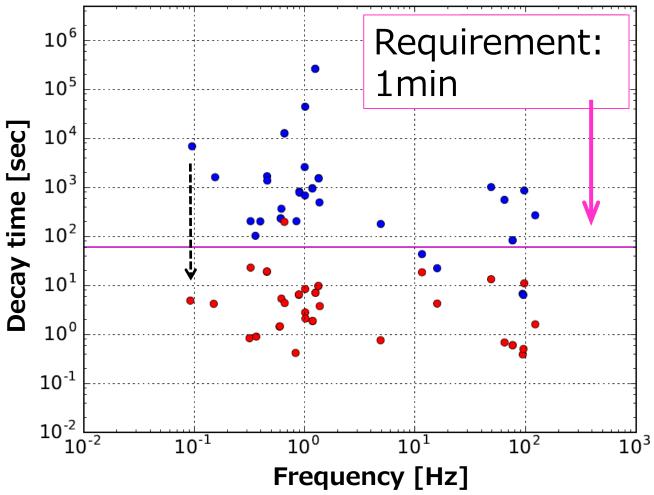




Calm-down phase:
Suppress large disturbance

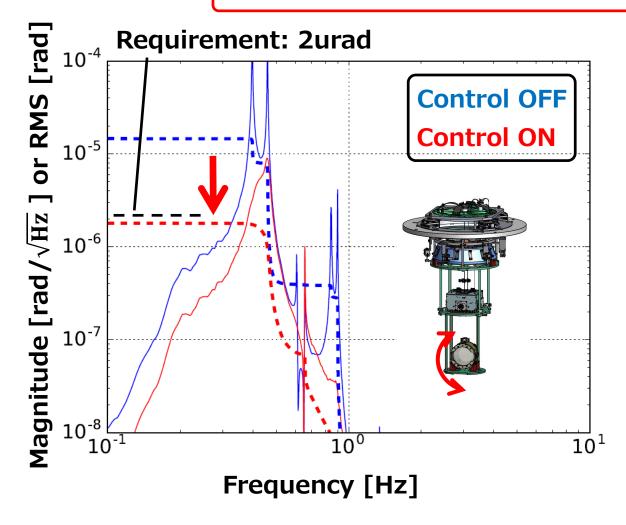
Control OFF Control ON





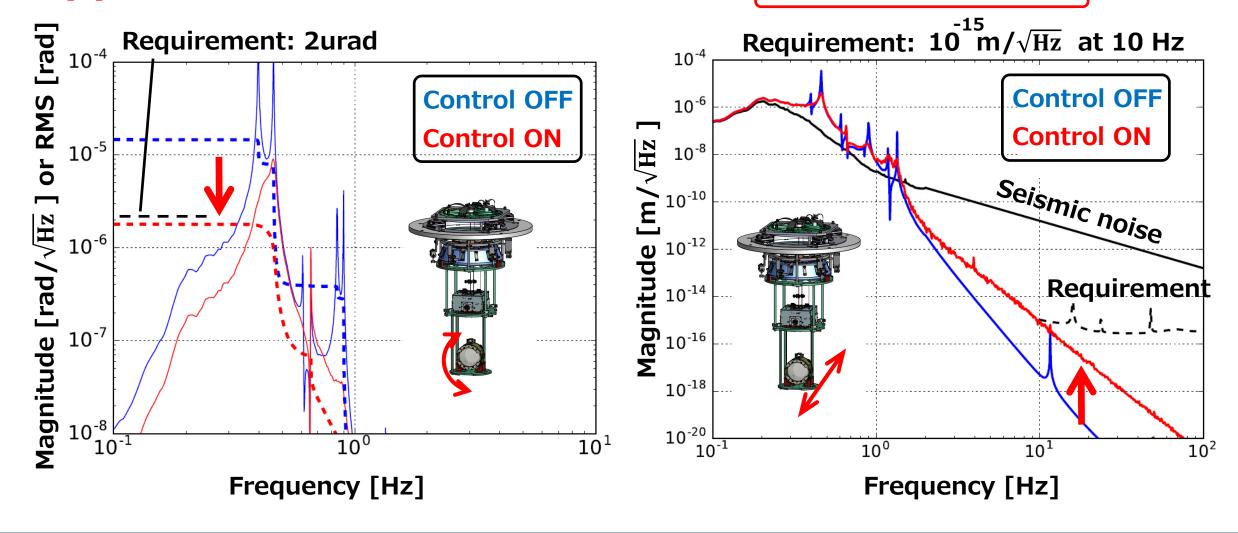
Observation phase:

Suppress RMS (Root Mean Square) & control noise



Observation phase:

Suppress RMS (Root Mean Square) & control noise



Summary 2

- 1) An iKAGRA suspension was assembled for iKAGRA operation.
- 2) Its performance were tested.
 - → Measurement: consistent with simulation.
- 3) Active control system for a suspension is designed.
 - → Clam-down phase: resonances → damped.
 - → Observation phase: RMS & control noise → suppressed.

Summary

1. Source localization

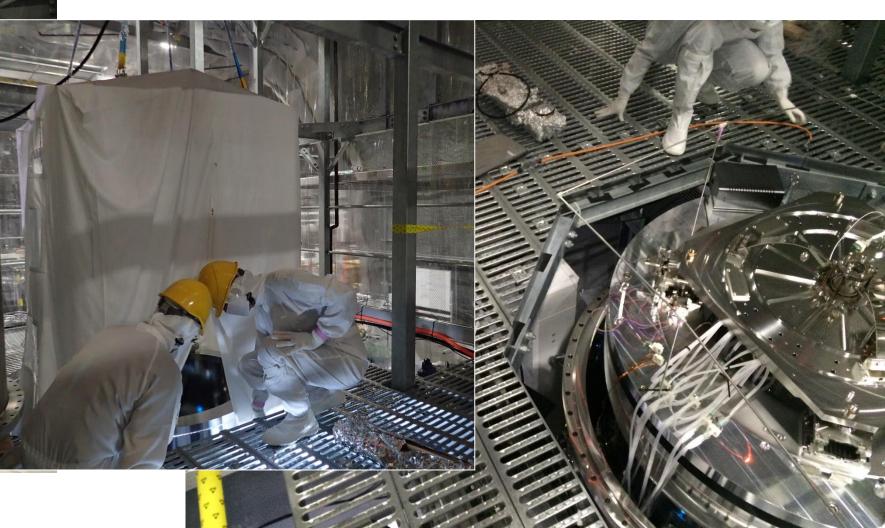
- A localization with hierarchical network is demonstrated.
- > Low sensitivity detector can contribute.
 - → 4th detector contributes. → useful for follow-up observation.

2. Detector development

- 1) iKAGRA-PR3 suspension was assembled for iKAGRA operation.
- 2) Its performance were tested.
 - > Simulation was consistent with measurement.
 - 3) Active control system for type-Bp suspension is designed
 - → Clam-down phase: resonances → damped.
 - → Observation phase: RMS & control noise → suppressed.

Back up

Modern NINJAs in the Kamioka mine.



Summary

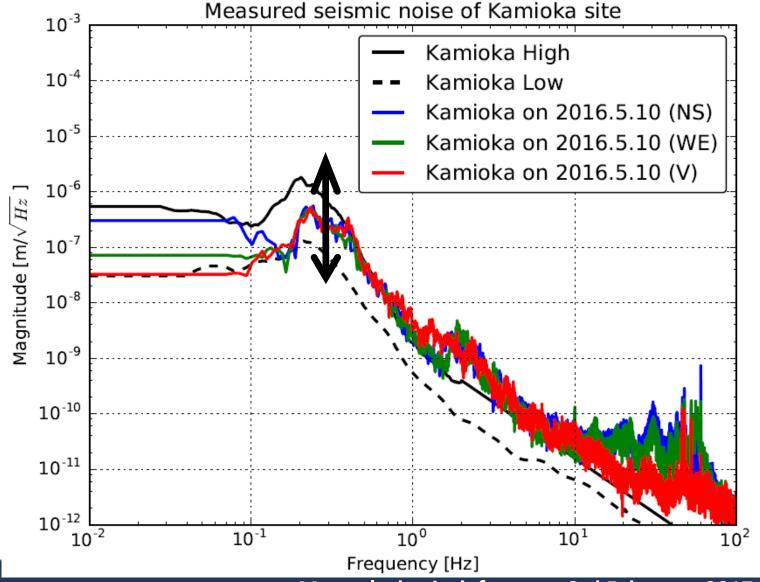
1. Source localization

- A localization with hierarchical network is demonstrated.
- → Low sensitivity detector can contribute.
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2. Detector development

- 1) iKAGRA-PR3 suspension was assembled for iKAGRA operation.
- 2) Its performance were tested.
 - > Simulation was consistent with measurement.
- 3) Active control system for type-Bp suspension is designed.
 - → Clam-down phase: resonances → damped.
 - → Observation phase: RMS & control noise → suppressed.

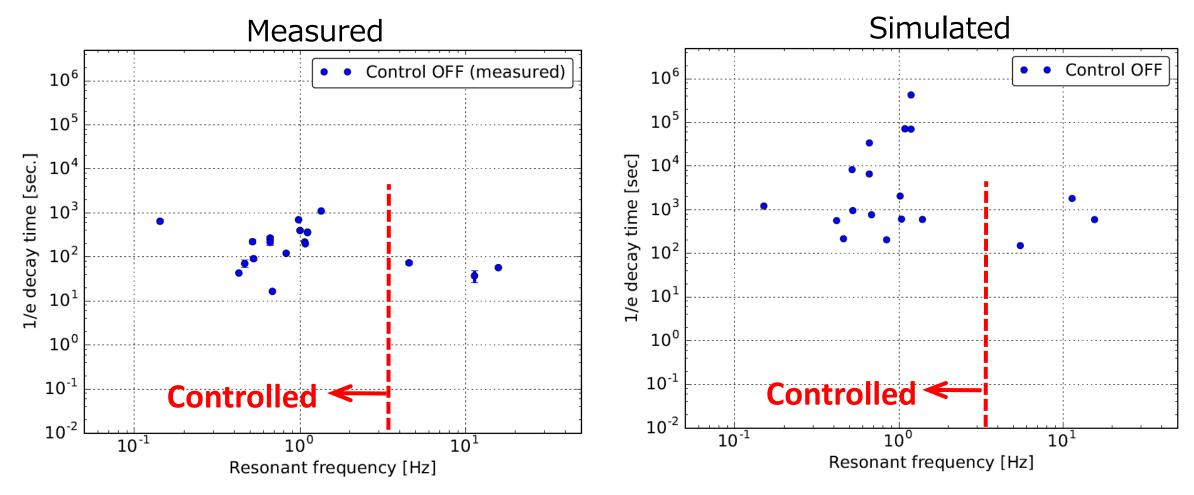
Seismic noise of Kamioka (on 2016.5.10)



seismic noise was measured on 2016.5.10.

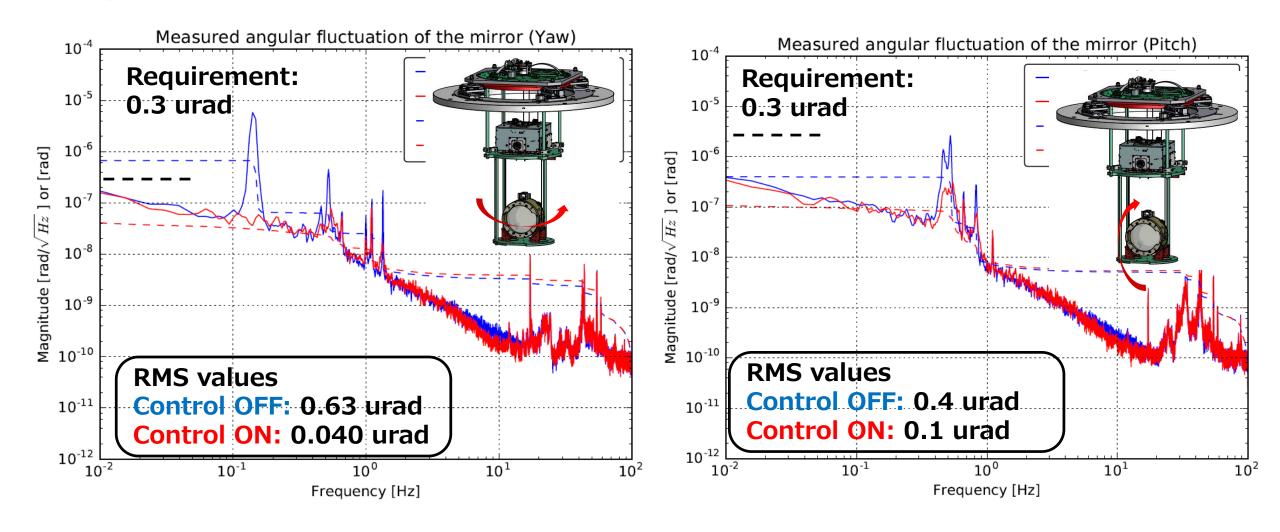
PR3 measurement was conducted on 2016.5.24.

Performance test 1

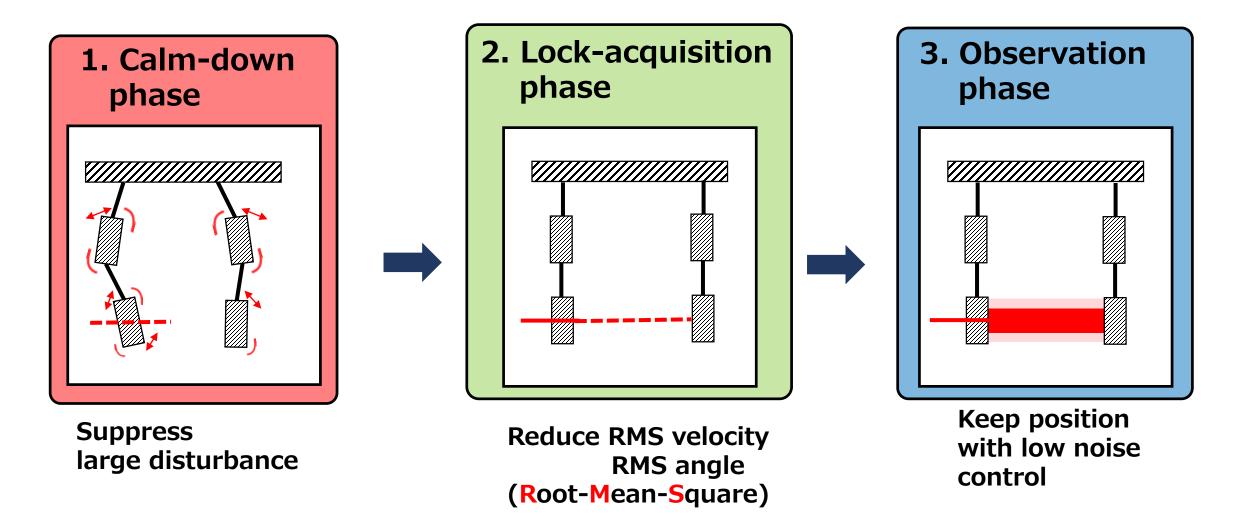


1) Control OFF → Necessary to feedback measurement.

Angular fluctuation of the mirror



Designing active control system / Control phase

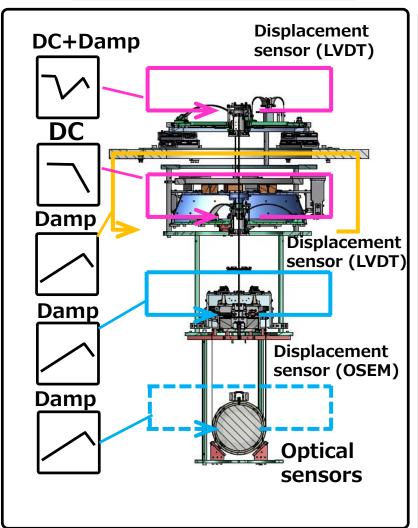


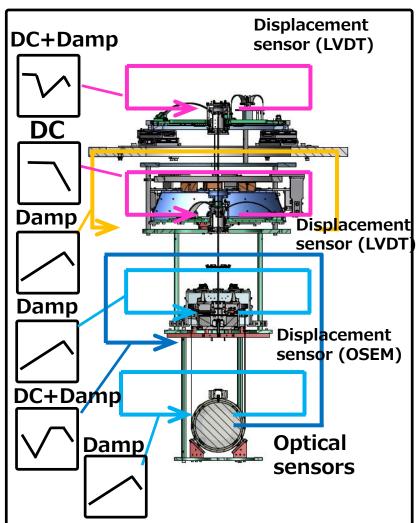
Designing active control system / Type-Bp SAS

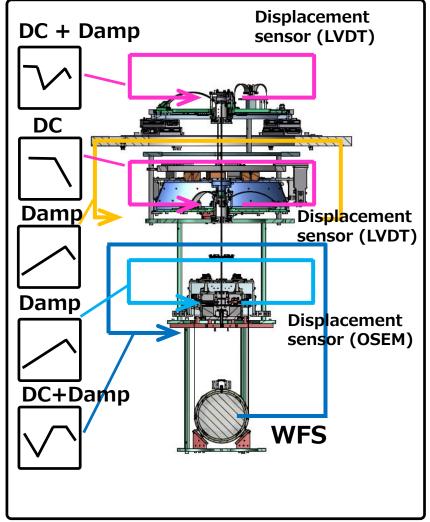
1. Calm-down phase

2. Lock-acquisition phase

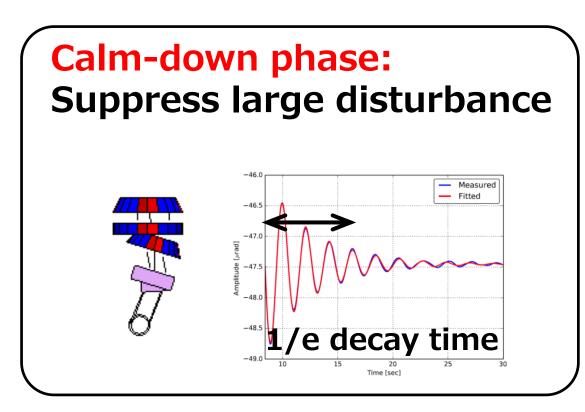
3. Observation phase

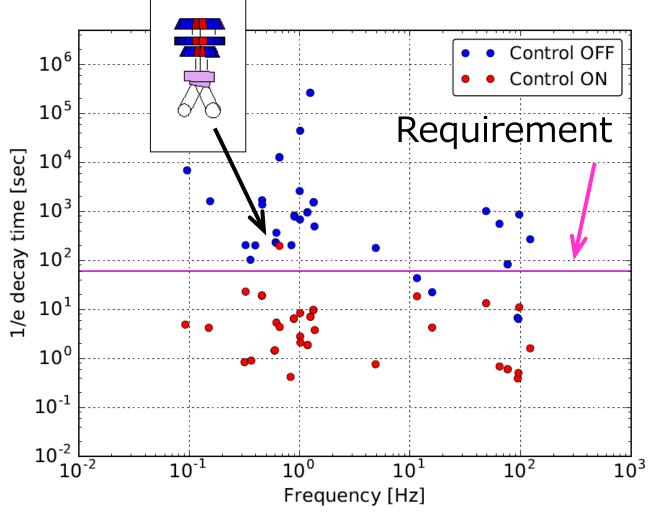


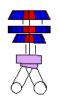




Designing active control system 1



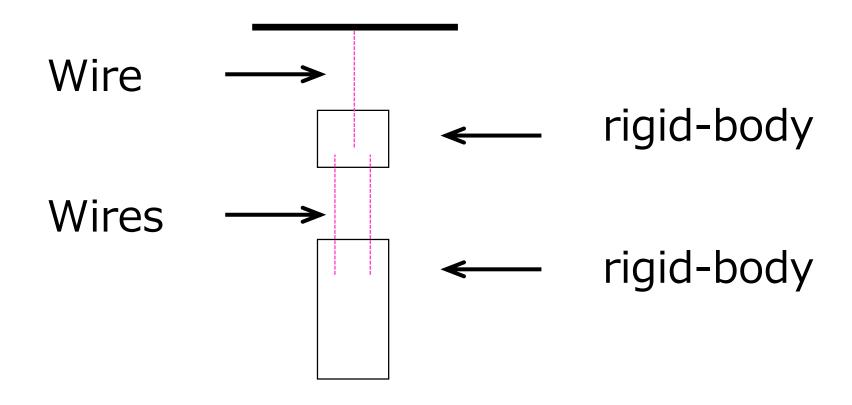




Not disturb operation → No problem.

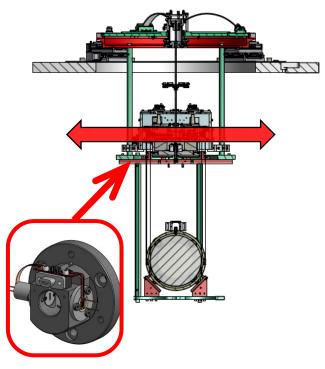
(if all sensors available)

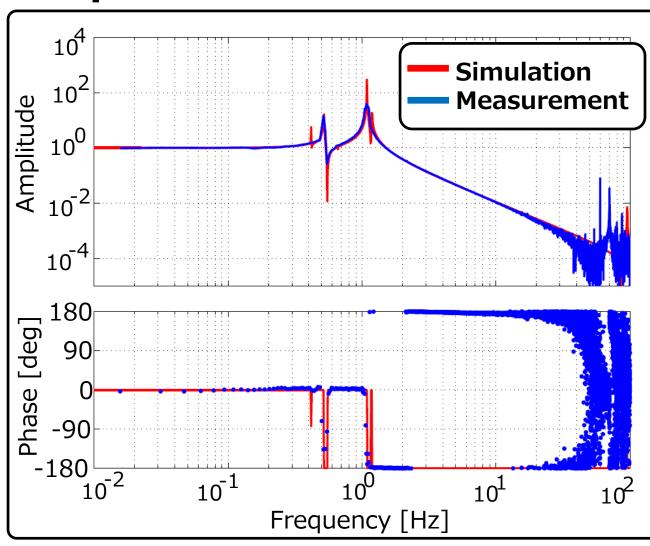
Simulation model: Based on rigid-body



1. Frequency response

For each Components:



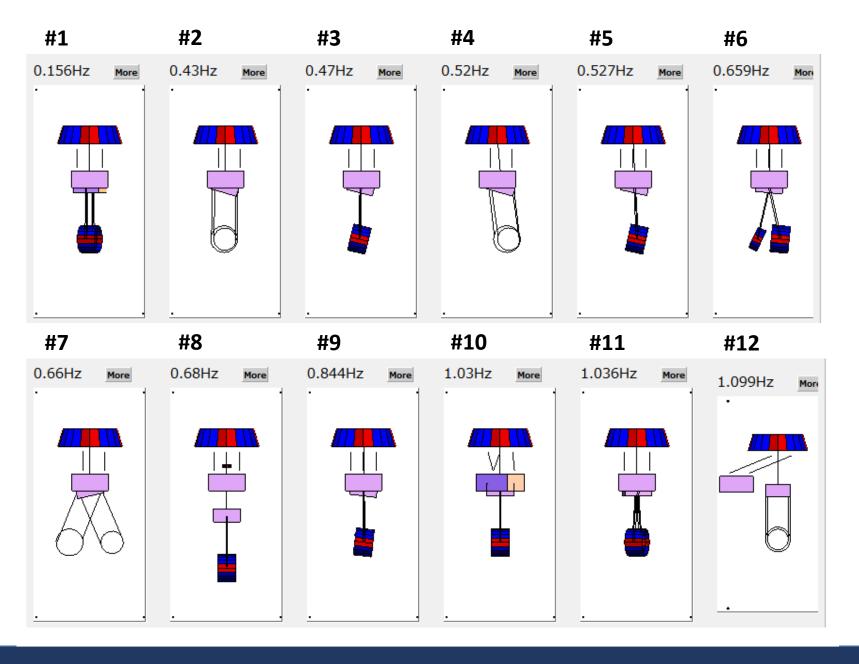


Band width: 0.02 Hz

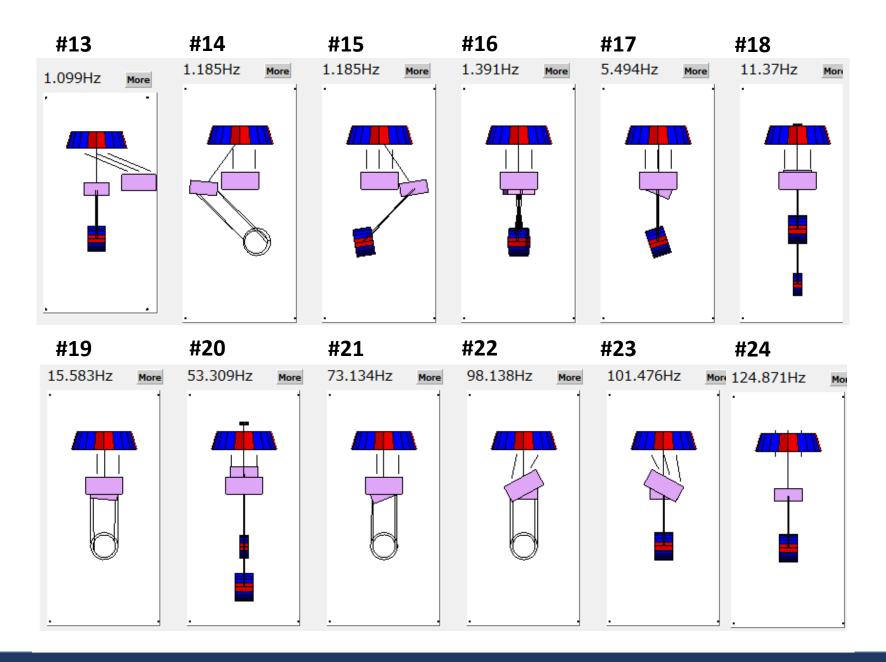
Window: Hanning

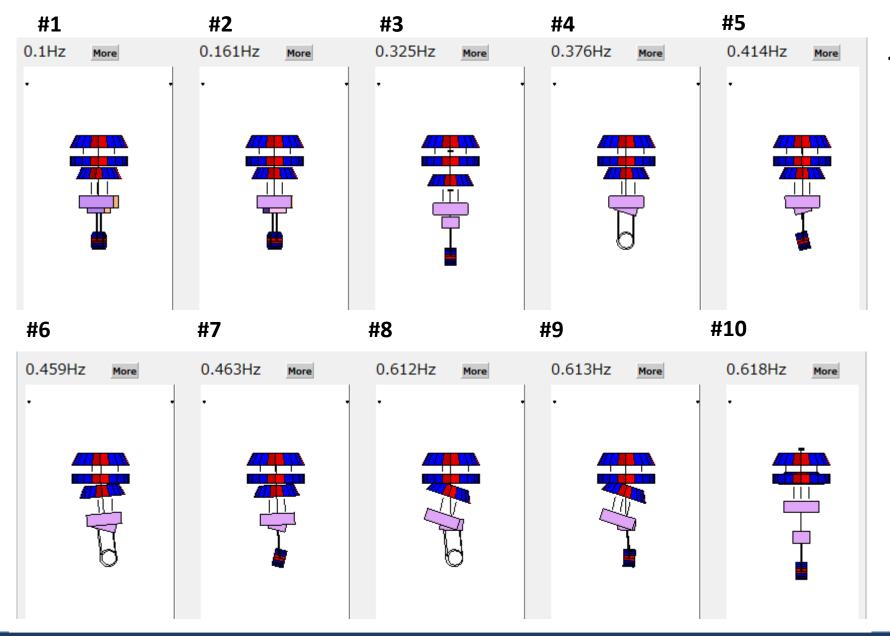
Overlap: 50 %

Average: 5

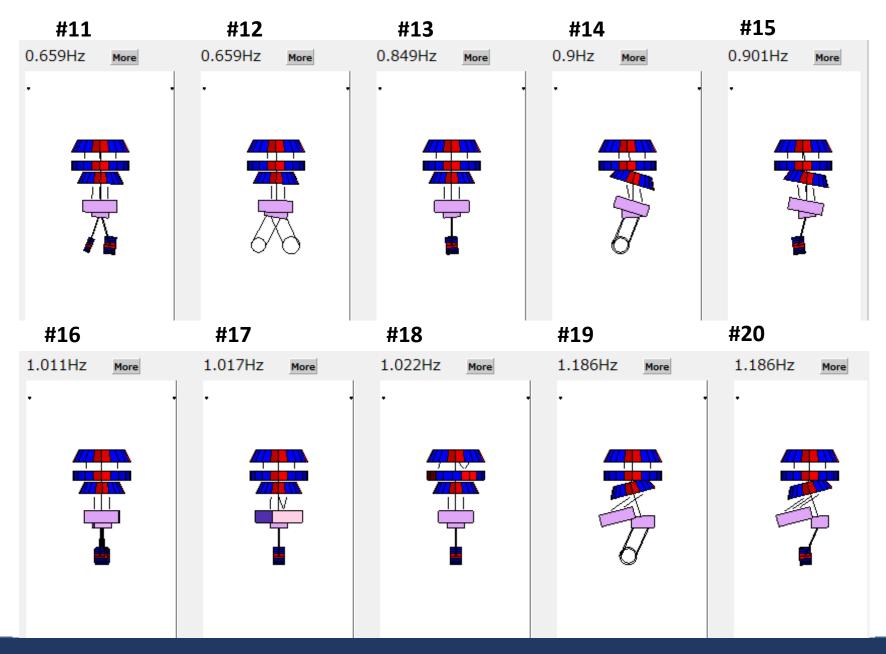


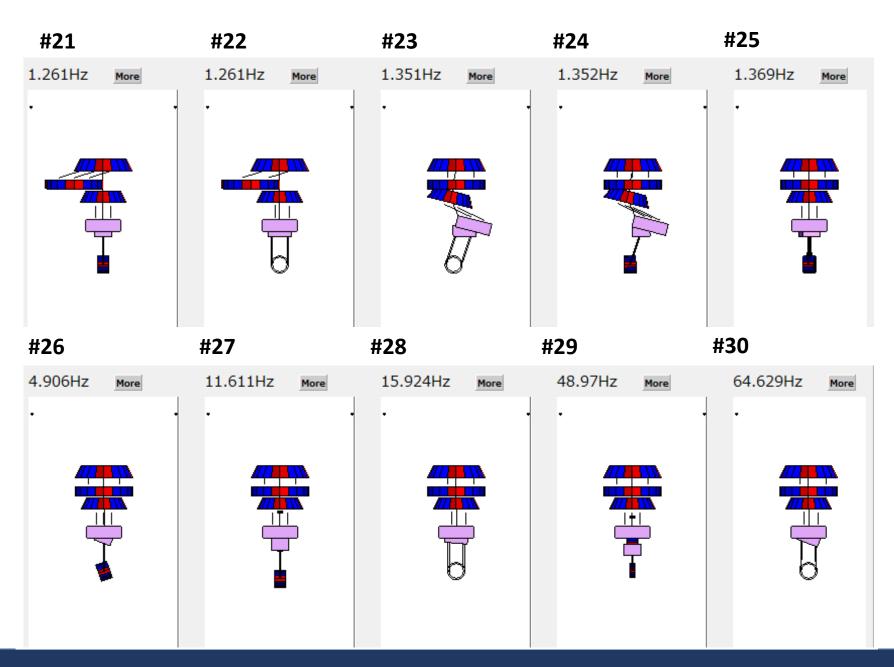
TypeBpp SAS
Eigen mode List: 24 modes

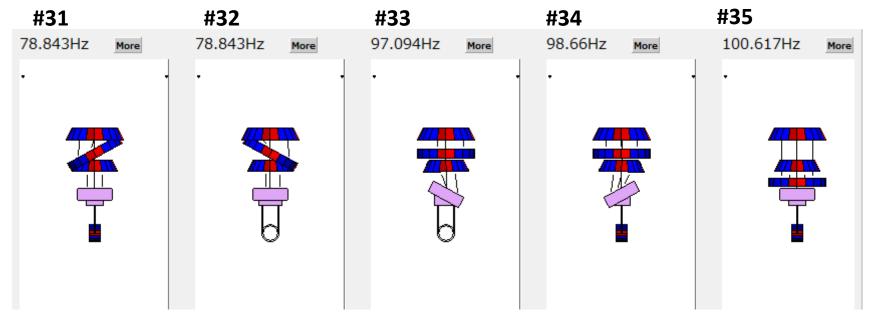




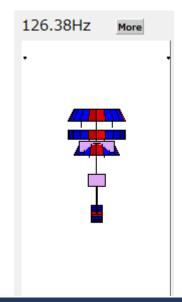
TypeBp SAS
Eigen mode List: 36 modes

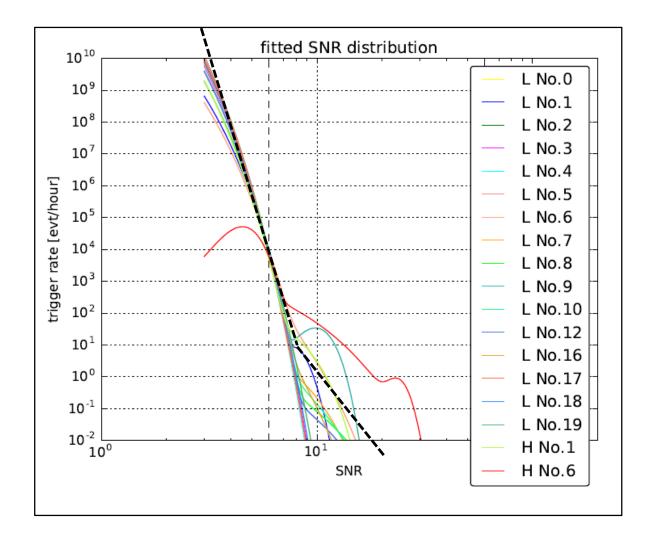




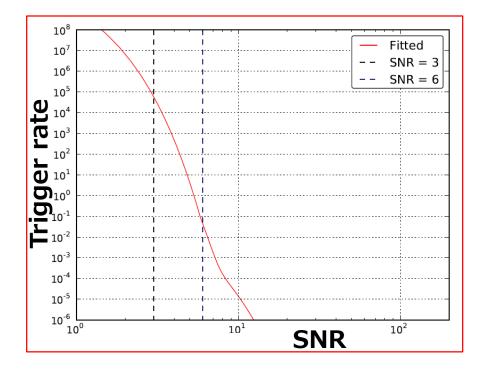


#36









Calculation setup: False Alarm Probability (FAP)

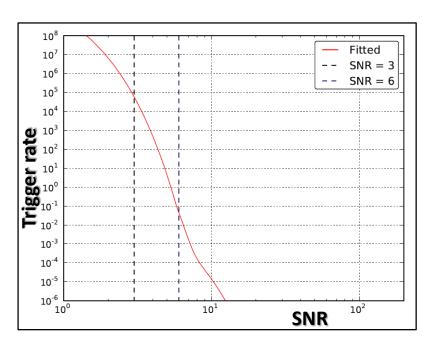
SNR distribution (per template)

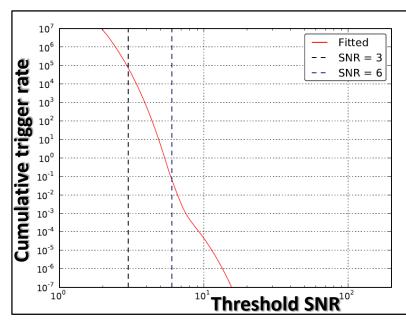


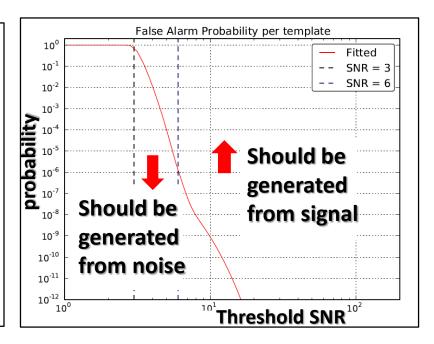
Cumulative SNR distribution (per template)



False Alarm Probability (per template)







$$FAP = 1 - \exp(-R \times T)$$

R = cumulative rate of background triggers per template,
 above a given threshold, per template,
 T = analyzing time for the V1 (less sensitive detector)

Calculation setup / 3 detector network by HLV

2. Transform HL into *HLV* coincidences.

1) Generating V1 triggers

V1 trigger based on random parameters : Vr (from noise)

SNR = random following measurement **Timing** = tH1 or tL1 + random [-35ms:35ms] **Phase** = random [0:2 π]

V1 trigger based on injection parameters : Vi (from signal)

SNR = metadata + Gauss(0,1) Timing = metadata + Gauss(0,0.66 ms* $\frac{6}{SNR}$) Phase = measured + Gauss(0,0.25 rad)

2) Mixing V1 triggers

Case 1: worst case HL+Vr, or HL

(Based on *FAP*)

Case 2: best case HL+Vi, or HL
(Based on *SNR*th)

Case 3: Realistic case HL+Vr, or HL+Vi, or HL (Based on *FAP* and *SNR*th)

Mixing triggers: HL → HL or HLV

Considered 3 patterns:

```
Case 1: HL \rightarrow HL or HL + random V

If p > FAP, otherwise

Noise-based trigger, or HL

Case 2: HL \rightarrow HL, or HL + V based on injection

If V1 \text{ SNR} < \text{threshold}, otherwise

Signal-based trigger, or HL

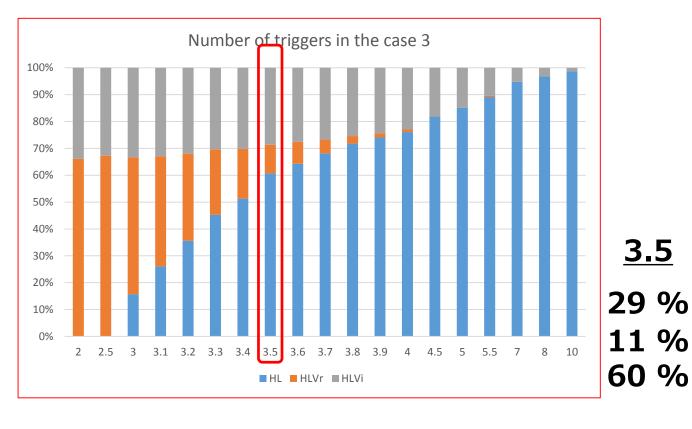
Case 3: HL \rightarrow HL, or HL + random V, or HL + V based on injection
```

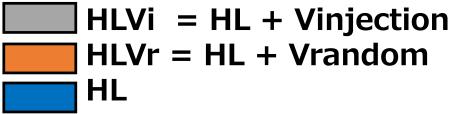
If p > FAP and V1 SNR < threshold, If p < FAP, If p > FAP and V1 SNR > threshold

- Noise-based trigger, or
 signal-based trigger, or
 HL

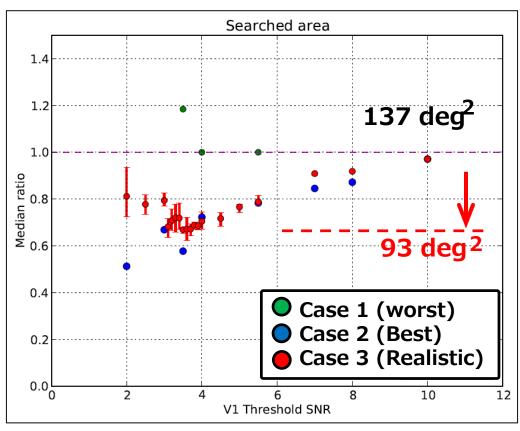
Update the sky localization performance in the case 3: Summary of sky localization performance



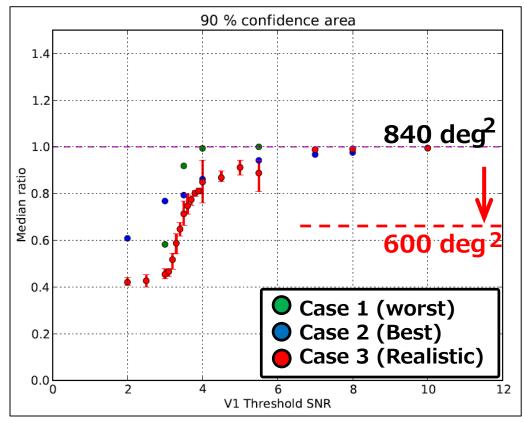




Optimization of Virgo threshold:

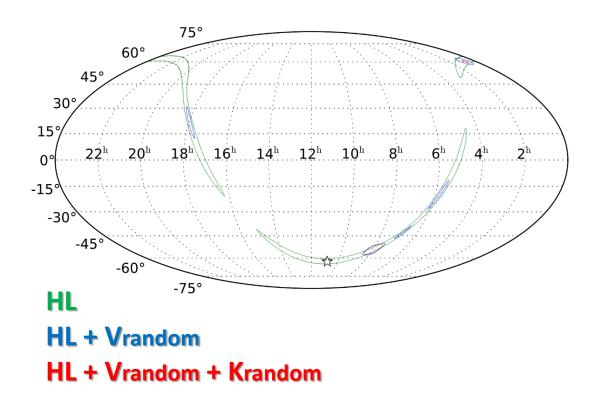


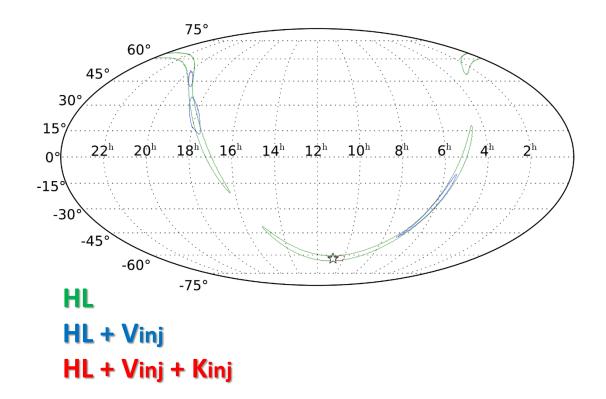
SNR threshold for H, L = 5.



SNR threshold for H, L = 5.

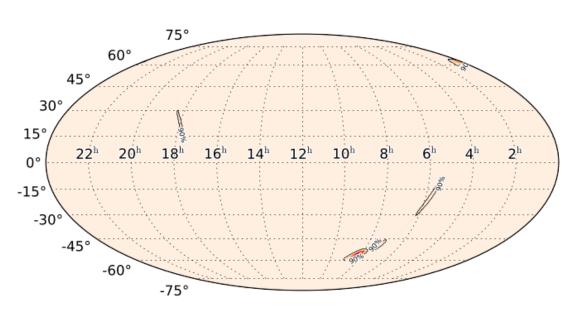
* Start to generate skymaps with 4 detector (V1, K1 threshold = 3.5)





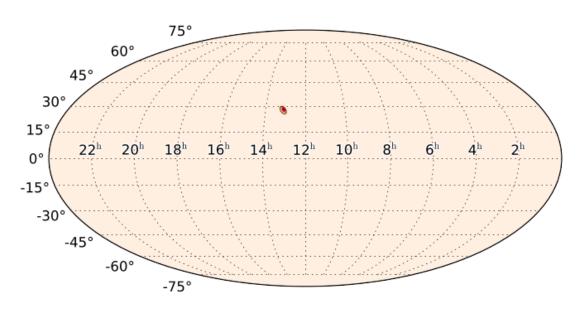
Expected localization performance / by HLV

HL+Vrandom



SNR (H)	SNR (L)	SNR(V)
12.8	11.5	4.5

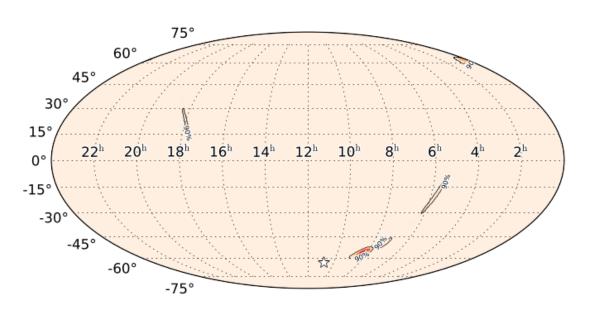
HL+Vinjection



SNR (H)	SNR (L)	SNR(V)
16.5	17.1	3.9

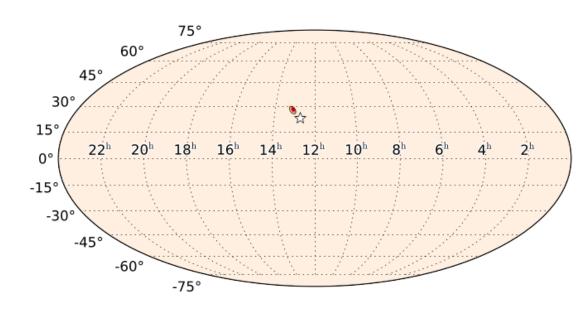
Expected localization performance / by HLV





SNR (H)	SNR (L)	SNR(V)
12.8	11.5	4.5

HL+Vinjection

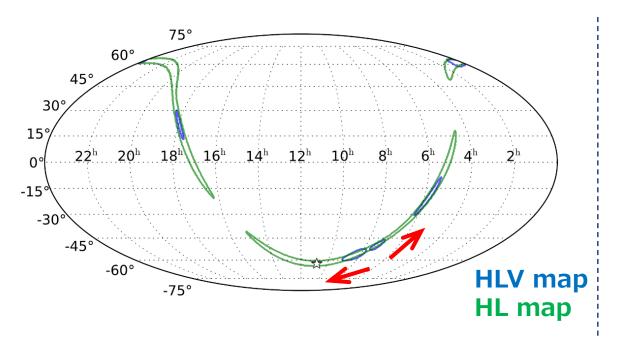


SNR (H)	SNR (L)	SNR(V)
16.5	17.1	3.9

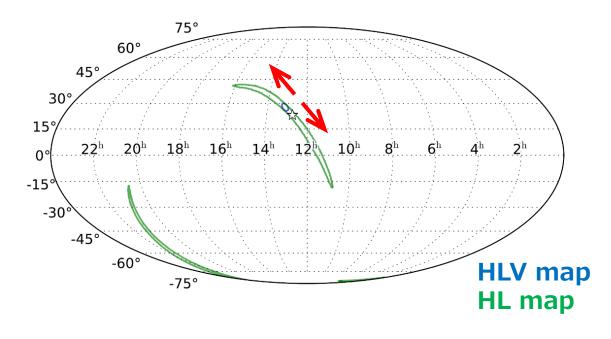
- Typical sky maps in this method
 - → sometimes fail to predict the location within 90 % confidence area.

Expected localization performance / by HLV

HL+Vrandom



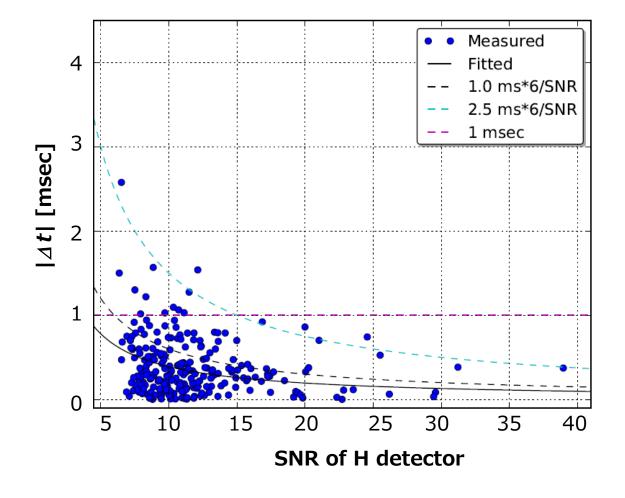
HL+Vinjection



- In this hierarchical network search,
 HLV sky map → If there is no EM-counterpart in HLV map, HL map.
- It will be useful for GW-EM follow-up observation.

For further accuracy improvement:

Measured uncertainties on arrival time vs. SNR.



Relation between timing error and SNR

Detected arrival timing has some uncertainties Δt due to:

- 1) calibration uncertainty
- 2) discrepancies of templates. and so on.

If SNR becomes large, Δt becomes small.

Since, accuracy largely depends on Δt ,

For further improvement of accuracy,

- → Necessary to reduce timing error
- → Necessary to improve sensitivity of GW detectors.

Calculation setup / 4 detector network by HLVK

2. Transform HL into *HLVK* coincidences.

1) Generating V1 triggers

V1 trigger based on random parameters : Vr, Kr

SNR = random following measurement Timing = tH1 or tL1 + random [-35ms:35ms] Phase = random $[0:2\pi]$

V1 trigger based on injection parameters : Vi, Ki

SNR = metadata + Gauss(0,1) Timing = metadata $+ Gauss(0,0.66 ms*\frac{6}{SNR})$ Phase = measured + Gauss(0,0.25 rad) 2) Mixing V1 triggers

Case 1: worst case
HL+Vr, HL+Kr, HL+Vr+Kr or HL
(Based on *FAP*)

Case 2: best case
HL+Vi, HL+Ki, HL+Vi+Ki or HL
(Based on *SNR*th)

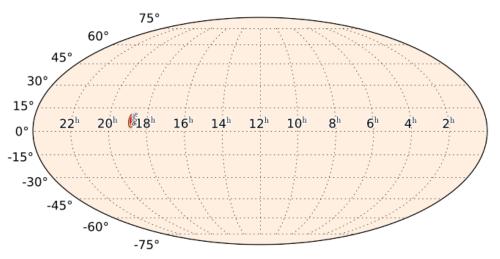
Case 3: Realistic case
HL+Vr, HL+Kr, HL+Vr+Kr,
HL+Vi, HL+Ki, HL+KVi+Ki,
HL+Vr+Ki, HL+ViKr, or HL

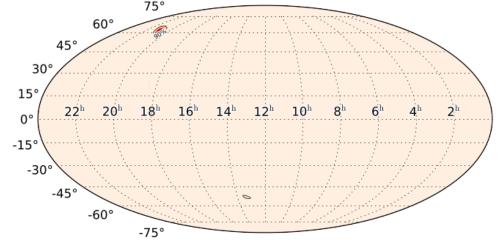
(Based on *FAP* and *SNR*th)

Expected localization performance / by HLVK



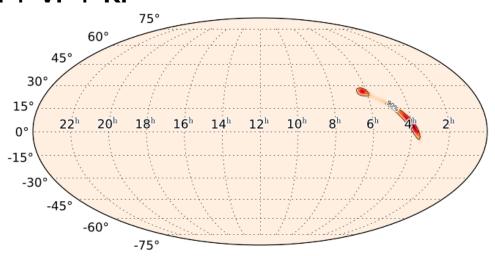


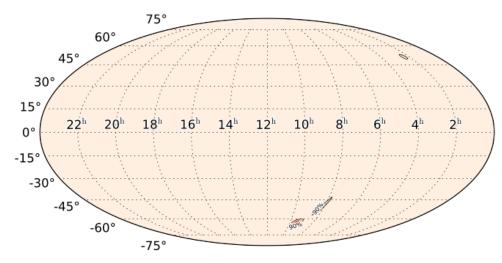






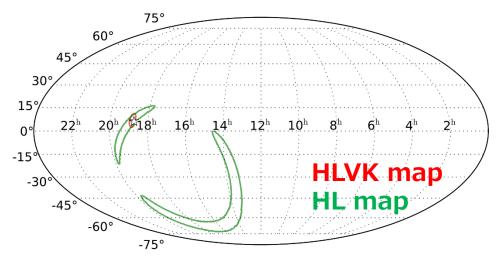
HL + Vi + Kr



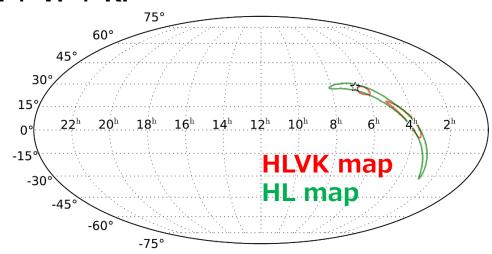


Expected localization performance / by HLVK

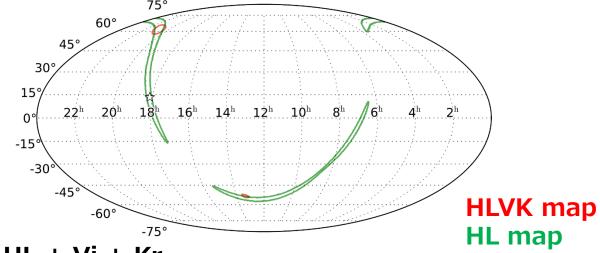




HL + Vr + Ki



HL + Vr + Kr



HL + Vi + Kr

