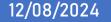


# Finding working point of an interferometer with Finesse 3

#### Enzo Tapia S. for the Finesse team

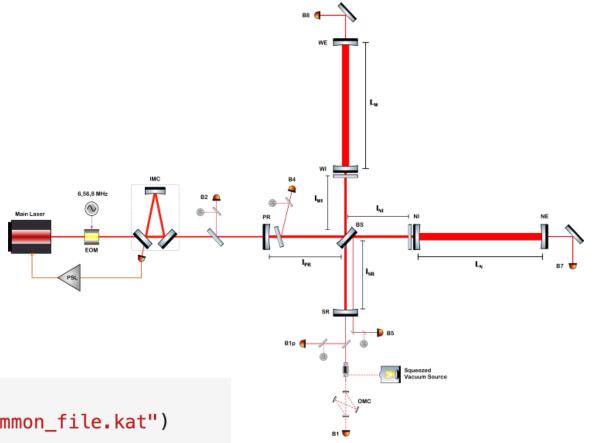


Working point of detector with Finesse 3

### A model of the detector:

Recommendation: Make a drawing of what you'd like to model.

Then write or import a kat-script that represents the interferometer we want to simulate.



model0 = finesse.Model()
model0.parse\_file("Virgo\_local\_katscript/00\_virgo\_common\_file.kat")
model1 = model0.deepcopy()
model1.parse\_file("Virgo\_local\_katscript/01\_additional\_katscript.kat")

### **Check and modify parameters:**

<pre>model1.i1. model1.i1.</pre>	P = 25 parameter_t
Description	Value
ower	25.0 W
Phase	0.0 degrees
Frequency	0.0 Hz
Signals only	False

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### **Adjust recycling cavities**

## Adjust recycling cavity lengths to fulfil requirement (design).

#### model1.get('lPOP\_BS').L

<lp><lPOP\_BS.L=5.9399 @ 0x15dcfbf40>

adjust\_recycling\_cavity\_length(model1, "PRC", "lPRC", "lPOP\_BS", verbose = True)

 adjusting PRC length adjusting lPOP\_BS.L by 0.0004736 m

model1.get('lPOP\_BS').L

<lPOP\_BS.L=5.94037359652047 @ 0x15dcfbf40>

model1.get('lsr').L

<lsr.L=5.943 @ 0x15ddf9a80>

adjust\_recycling\_cavity\_length(model1, "SRC", "lSRC", "lsr", verbose = True)

 adjusting SRC length adjusting lsr.L by 0.000883 m

model1.get('lsr').L

<lsr.L=5.94388296505047 @ 0x15ddf9a80>

def adjust\_recycling\_cavity\_length(

model, cavity: str, L\_in: str, S\_out: str, verbose=False
):

"""Adjust cavity length so that it fulfils the requirement:

L = 0.5 \* c / (2 \* f6), see TDR 2.3 (VIR-0128A-12).

#### Parameters

#### cavity : str

Name of the cavity being adjusted.

#### L\_in : str

Variable used to define the length of the cavity.

Needed because the common file does not use a variable. S\_out : str

Name of the space component used to adjust the cavity.

# works also for legacy
f6 = model.get("eom6.f").value

if verbose:
 print(f"- adjusting {cavity} length")

# calculate the required adjustment
tmp = 0.5 \* CONSTANTS["c0"] / (2 \* f6)
delta\_l = tmp.eval() - model.get(L\_in).value.eval()

if verbose: print(f" adjusting {S\_out}.L by {delta\_l:.4g} m")

# apply the adjustment
model.get(S\_out).L += delta\_l

### **Check sensitivity:**

def get\_QNLS(model, axis=[5, 5000, 100]):
 kat = model.deepcopy()
 kat.parse(
 """#kat
 # Differentially modulate the arm lengths
 fsig(1)
 sgen darmx LN.h
 sgen darmy LW.h phase=180
 # Output the full quantum noise limited sensitivity
 qnoised NSR\_with\_RP B1.p1.i nsr=True
 #qnoised NSR\_with\_RP SR.p2.o nsr=True

```
# Output just the shot noise limited sensitivity
  qshot NSR_without_RP B1.p1.i nsr=True
```

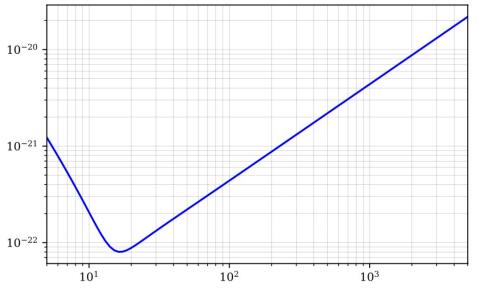
return kat.run(f'xaxis(darmx.f, "log", {axis[0]}, {axis[1]}, {axis[2]})')

But before this step, it is important to set the parameter of propagation of the fundamental mode to 'False'.

model1.\_settings.phase\_config.zero\_k00=False

out\_sensitivity1 = get\_QNLS(model1)
plt.loglog(out\_sensitivity1.x1, np.abs(out\_sensitivity1['NSR\_with\_RP']))
range\_bns1 = inspiral\_range.range(out\_sensitivity1.x1, np.abs(out\_sensitivity1['NSR\_with\_RP'])\*\*2)
print('BNS range:', np.round(range\_bns1, 3), 'Mpc')

BNS range: 17.262 Mpc



Note: Here we are experimenting also with the 'inspiral\_range' module from Pygwinc. https://git.ligo.org/gwinc/inspiral-range

.....

### **Check sensitivity:**

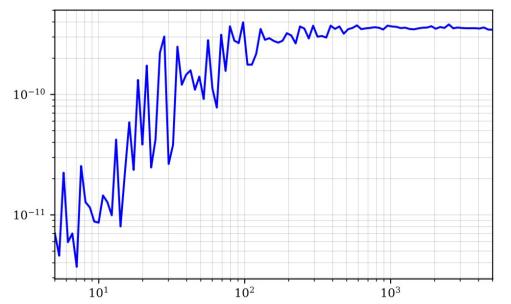
But before this step, it is important to set the parameter of propagation of the fundamental mode to 'False'.

The models have this parameter as 'True' by default, thus we get an unwanted working point with a low sensitivity. model1.\_settings.phase\_config.zero\_k00

True

out\_sensitivity1 = get\_QNLS(model1)
plt.loglog(out\_sensitivity1.x1, np.abs(out\_sensitivity1['NSR\_with\_RP']))
#range\_bns1 = inspiral\_range.range(out\_sensitivity1.x1, np.abs(out\_sensitivity1['NSR\_with\_RP'])\*\*2)
#print('BNS range:', np.round(range\_bns1, 3), 'Mpc')

[<matplotlib.lines.Line2D at 0x2a006f760>]



# Finding and operating point: Pre-tuning

#### Following these steps:

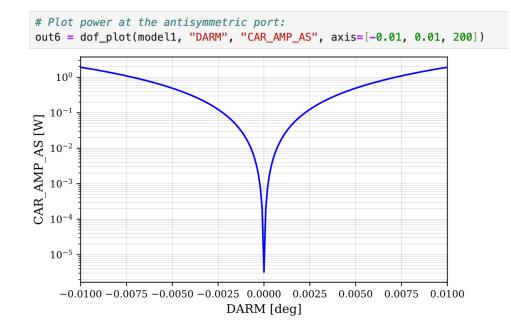
- Only use main carrier field in the following steps.
- Remove any cross-coupling between the two arms, so we can tune them independently from each other.
- Arm cavities are on resonance.
- Interferometer at the dark fringe.
- PRC is on resonance.
- SRC is on resonance.
- SRC is in the condition for Resonant Sideband Extraction (RSE).
- Restore the modulators.

```
## Run the model and make this changes in series:
model2 = model1.deepcopy()
out1 = model2.run(
    fac.TemporaryParameters(
        fac.Series(
            # Switch off the modulators and misalign SR and PR mirrors:
            fac.Change({"eom6.midx": 0, "eom8.midx": 0, "eom56.midx": 0,
                        "SR.misaligned": True, "PR.misaligned": True,
                        "SRAR.misaligned": True, "PRAR.misaligned": True}),
            # Maximize arm powers:
            fac.Maximize("B7_DC", "NE_z.DC", bounds=[-180,180], tol=1e-14),
            fac.Maximize("B8_DC", "WE_z.DC", bounds=[-180,180], tol=1e-14),
            # Minimize dark fringe power:
            fac.Minimize("B1_DC", "MICH.DC", bounds=[-180,180], tol=1e-14),
            # Align back PRM:
            fac.Change({"PR.misaligned": False}),
            # Maximize PRC power:
            fac.Maximize("CAR_AMP_BS", "PRCL.DC", bounds=[-180,180], tol=1e-14),
            # Align back SRM:
            fac.Change({"SR.misaligned": False}),
            # Maximize SRC power, then offset by 90 deg:
            fac.Change({"SRCL.DC": 0}),
            fac.Maximize("B1_DC", "SRCL.DC", bounds=[-180,180], tol=1e-14),
            fac.Change({"SRCL.DC": -90}, relative=True)),
        exclude=("PR.phi", "NI.phi", "NE.phi", "WI.phi", "WE.phi", "SR.phi",
                 "NE_z.DC", "WE_z.DC", "MICH.DC", "PRCL.DC", "SRCL.DC")
```

# Finding and operating point: Pre-tuning

#### Check each relevant figure of merit:

- Maximum power circulating in the arms.
- MICH tuned to dark fringe.
- PRCL tuned for maximum power inside the PRC.
- SRCL tuned for minimum power inside the SRC.
- DARM tuned for minimum power on B1.



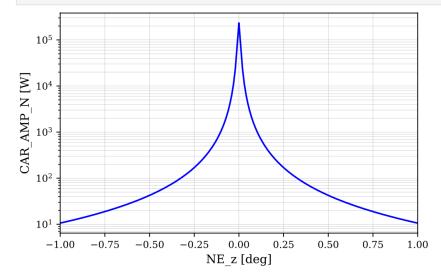
```
def dof_plot(model, dof, detector, axis=[-1, 1, 200], show=True):
    """Sweep across a DoF, reading out at the provided (amplitude) detector."""
    axis = np.array(axis, dtype=np.float64)
    #axis[:2] *= xscale
    temp = model
    out = temp.run(
        fac.Xaxis(f"{dof}.DC", "lin", axis[0], axis[1], axis[2], relative=True)
    )
```

plt.semilogy(out.x[0], (np.abs(out[detector]))\*\*2) # these are amplitude detectors.
plt.xlabel(model.get(dof+'.name') + " [deg]")
plt.ylabel(model.get(detector+'.name') + " [W]")

return out

#### # Plot power in North arm out1 = dof\_plot(model1, 'NE\_z', 'CAR\_AMP\_N')

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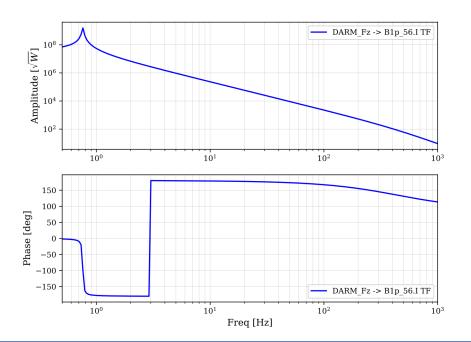


# Check sensitivity, plot DARM TF.

The sensitivity can be improved with a locking scheme and not only the pre-tuning process.

We can check the DARM TF too. However, we should be checking more figures of merit.

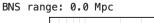
\* Other figures of merit? Green lights, sensitivity in Mpc.

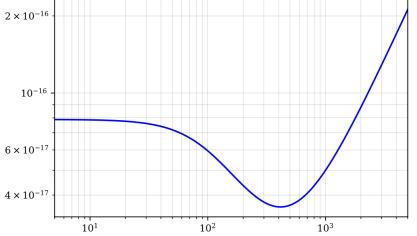


```
def get_QNLS(model, axis=[5, 5000, 100]):
    kat = model.deepcopy()
    kat.parse(
        """"#kat
        # Differentially modulate the arm lengths
        fsig(1)
        sgen darmx LN.h
        sgen darmy LW.h phase=180
        # Output the full quantum noise limited sensitivity
        qnoised NSR_with_RP B1.p1.i nsr=True
        # qnoised NSR_with_RP SR.p2.o nsr=True
        # Output just the shot noise limited sensitivity
        qshot NSR_without_RP B1.p1.i nsr=True
    """"
)
```

return kat.run(f'xaxis(darmx.f, "log", {axis[0]}, {axis[1]}, {axis[2]})')







### Identify and Optimize a Sensing matrix

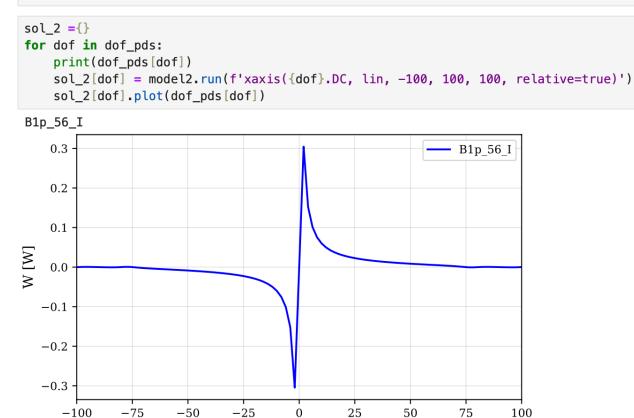
#### **Optimize demodulation phases:**

_										<pre>model to 4 = model3</pre>	•						
De	Define a sensing matrix:						<pre>for dof in dof_pds.keys():     model4.run(fac.OptimiseRFReadoutPhaseDC(dof, dof_pds[dof]))</pre>										
sens_m	natrix_dc	= model3.r	un(fac.se	ensing.Se	nsingMatr	ixDC(											
['DARM', 'CARM', 'MICH', 'PRCL', 'SRCL'], ['B1p_56', 'B2_6', 'B2_56', 'B2_8'] ))						sens_	<pre>sens_matrix_dc_4 = model4.run(fac.sensing.SensingMatrixDC(</pre>										
<pre># Display sensing matrix: sens_matrix_dc.display()</pre>					sens_												
Sells_I		urspray()								54 55 1							
	B1p_56_I	B1p_56_Q	B2_6_I	B2_6_Q	B2_56_I	B2_56_Q	B2_8_I	B2_8_Q		B1p_56_I	B1p_56_Q	B2_6_I	B2_6_Q	B2_56_I	B2_56_Q	B2_8_I	B2_8_Q
DARM	71	63	-0.014	0.0024	-0.0037	0.011	0.01	-0.00062	DARM	95	4.5E-08	-0.014	-1.5E-07	-0.011	0.0015	0.01	8.9E-09
CARM	0.18	0.16	3.8	-0.65	1	-3	-2.8	0.17	CARM	0.24	1.3E-06	3.9	-1E-07	3.1	-0.41	-2.8	-2.4E-06
місн	0.25	0.22	-1.2E-05	-4.4E-05	-0.00028	-0.00015	2.4E-05	-1.4E-06	МІСН	0.33	0.00011	-4E-06	-4.5E-05	5.1E-06	-0.00031	2.4E-05	3.6E-08
PRCL	0.00027	-0.00022	-0.0018	0.00033	0.00043	0.0024	0.0098	-0.00059	PRCL	5.5E-05	-0.00035	-0.0018	2.7E-05	-0.002	0.0015	0.0098	6.6E-10
SRCL	2.2E-05	1.1E-05	2.6E-06	-4.5E-07	7.4E-05	-0.00015	-2.7E-09	3.3E-10	SRCL	2.4E-05	-6.4E-06	2.6E-06	-1.3E-08	0.00016	-7.9E-09	-2.8E-09	1.6E-10

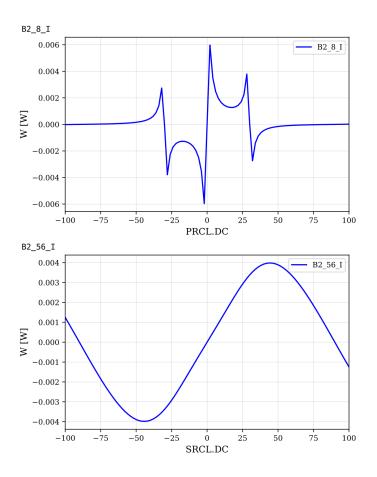
#### Check error signals

# Longitudinal DoF and readouts:

dof\_pds = {'DARM': 'B1p\_56\_I', 'CARM': 'B2\_6\_I', 'MICH': 'B2\_56\_Q', 'PRCL': 'B2\_8\_I',



DARM.DC



#### **Define a control scheme**

dof_pds
<pre>{'DARM': 'B1p_56_I',     'CARM': 'B2_6_I',     'MICH': 'B2_56_Q',     'PRCL': 'B2_8_I',     'SRCL': 'B2_56_I'}</pre>
<pre>model5 = model4.deepcopy() model5.parse(     """     # locks     lock DARM_lock B1p_56.outputs.I DARM.DC 0 1e-14     lock CARM_lock B2_6.outputs.I CARM.DC 0 1e-14     lock MICH_lock B2_56.outputs.Q MICH.DC 0 1e-11     lock PRCL_lock B2_8.outputs.I PRCL.DC 0 1e-12     lock SRCL_lock B2_56.outputs.I SRCL.DC 0 50e-11     # Add also RF and DC locks for DARM:     #lock DARM_rf_lock B1p_56.outputs.I DARM.DC 1 1e-14     lock DARM_dc_lock B1.outputs.DC DARM.DC 1 1e-14 offset=4m enabled=false     """)</pre>

Define the locking scheme for each DoF.

Note that this step is performed in the initialization of the Virgo model when using the finesse-virgo package.

### **Optimize lock gains**

Here we compute the gain of the PDH signal at the DC value of the TF from the DoF motion to the PDH response.

Then, we adjust the gain to be the reciprocal of the optical gain at DC. ## Optimize the gain of the lock: pdh\_gain = {} lock\_gain = {} for dof in dof\_pds\_I\_Q: pdh\_gain[dof] = np.abs(sol\_5[dof][dof\_pds\_I\_Q[dof], dof][0]) # Gain of the PDH signal in W/m lock\_gain[dof] = 1/pdh\_gain[dof] # Optimal gain of the lock lock\_gain[dof] \*= 2\*np.pi/ model5.lambda0\*np.rad2deg(1) # scaling the gain from m/W to deg/W #model5.get(f'{dof}\_lock.gain') = lock\_gain[dof] model5.set(f'{dof}\_lock.gain', lock\_gain[dof]) print(f'{dof} gain: ', model5.get(f'{dof}\_lock.gain')) DARM gain: 0.010480253074797585 CARM gain: 0.2606994562525331

MICH gain: 3653.7214307274107 PRCL gain: 98.1426349359976 SRCL gain: 6162.736040814418

#### Now we can run the locks.

### **Run locks**

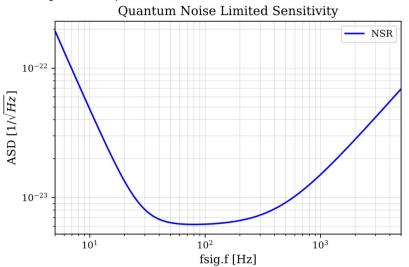
We run the locks with the current readout of DARM.

#### \* Other figures of merit? Green lights, sensitivity in Mpc...

virgo.plot\_QNLS()
out4 = virgo.get\_QNLS()
range bms4 = inspiral range range(out4 v1 m

range\_bns4 = inspiral\_range.range(out4.x1, np.abs(out4['NSR\_with\_RP'])\*\*2)
print('BNS range:', np.round(range\_bns4, 1), 'Mpc')

BNS range: 162.9 Mpc



model5.run(fac.RunLocks(method='newton'))

DARM\_lock < CARM\_lock < MICH\_lock < PRCL\_lock < SRCL\_lock < |

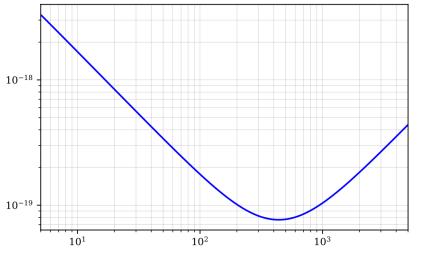
| 9997/10000

<RunLocksSolution of run locks @ 0x2914659a0 children=0>

Now the system is adjusted to an operating point (Heterodyne up to now), defined as the point where the PDH error signals are zero. Now let us check the figures of merit again.

out\_sensitivity5 = get\_QNLS(model5)
plt.loglog(out\_sensitivity5.x1, np.abs(out\_sensitivity5['NSR\_with\_RP']))
range\_bns5 = inspiral\_range.range(out\_sensitivity5.x1, np.abs(out\_sensitivity5['NSR\_with\_RP'])\*\*2)
print('BNS range:', np.round(range\_bns5, 3), 'Mpc')

BNS range: 0.005 Mpc



### **Change DARM from RF to DC readout**

These steps are equivalent to DARM\_RF\_to\_DC(dc\_offset = 0.005)

from the finesse-virgo package.

model6 = model5.deepcopy()
# Check RF locking:
model6.DARM\_lock.enabled.value

True

## Turn off the RF readout: model6.DARM\_lock.enabled = False # Check the DC value of DARM: print(model6.DARM.DC)

-1.1184696778681427e-07

# kick lock away from zero tuning for DC lock to grab with dc\_offset = 0.005 model6.DARM.DC += dc\_offset print(model6.DARM.DC)

0.004999888153032214

# Check current gain and
print(model6.DARM\_dc\_lock.gain)

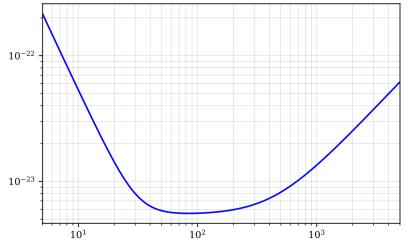
1.0

# take a guess at the gain lock\_gain=-0.01 model6.DARM\_dc\_lock.gain = lock\_gain model6.DARM\_dc\_lock.enabled = True

#### Check if this step gives us a good sensitivity level.

out\_sensitivity6 = get\_QNLS(model6)
plt.loglog(out\_sensitivity6.x1, np.abs(out\_sensitivity6['NSR\_with\_RP']))
range\_bns6 = inspiral\_range.range(out\_sensitivity6.x1, np.abs(out\_sensitivity6['NSR\_with\_RP'])\*\*2)
print('BNS range:', np.round(range\_bns6, 3), 'Mpc')

BNS range: 169.139 Mpc



\* Other figures of merit when designing? Green lights, sensitivity in Mpc...

### **Compute TF and cross couplings.**

We can compute these TFs with the FrequencyResponse() from finesse analysis actions.

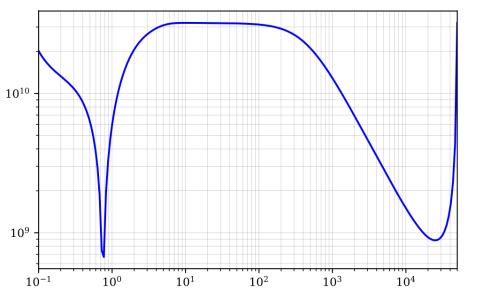
Note that its usage has changed.

plt.loglog(sol7.f, np.abs(sol7['DARM', 'B1p\_56.I']))

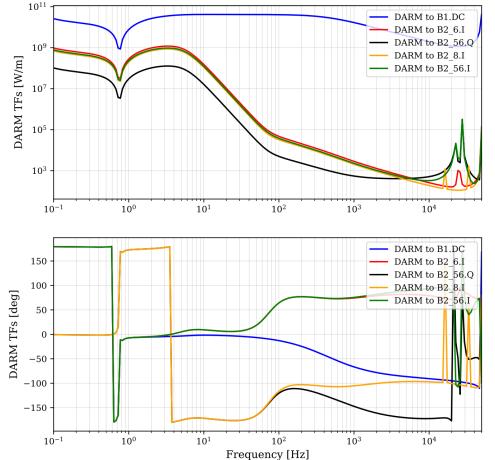
/var/folders/vf/trr1b62j04x\_tczb622t\_vzh0000gn/T/ipykernel\_34840/4274949489.py:1: DeprecationWarnin
g: FrequencyResponseSolution has changed to use [output, input], you seemed to have used [input, out
put] so returning that.

plt.loglog(sol7.f, np.abs(sol7['DARM', 'B1p\_56.I']))

[<matplotlib.lines.Line2D at 0x291b563e0>]

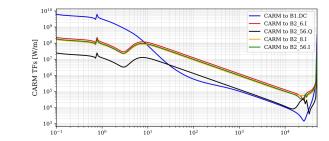


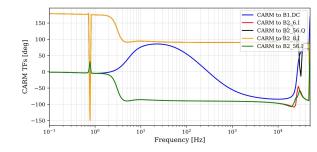
### **Compute TF and cross couplings.**

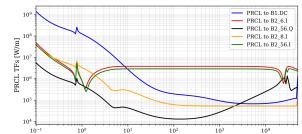


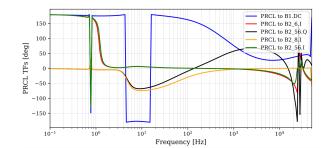
#### **Repository**:

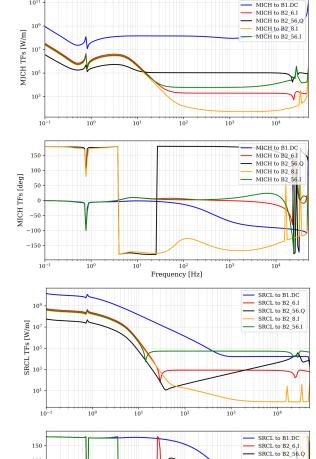
https://git.ligo.org/virgo/isc/finesse/enzo\_tapia/ -/tree/main/2024/LSC\_cross-couplings

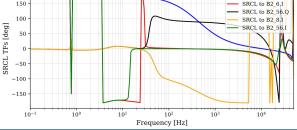












#### 06/07/2021

Working point of detector with Finesse 3