

Characterization of Depleted Monolithic Active Pixel Sensor (DMAPS) Prototypes

TWEPP 2014
Topical Workshop on
Electronics for Particle Physics

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Motivation: Pixel trackers for HEP experiments



Pile up of interactions at High Luminosity LHC

- Increased radiation dose
- More pile up

Goal: Keep pixel tracker performance by using improved concepts!

	STAR	ALICE-LHC	ILC	ATLAS-LHC	ATLAS-HL-LHC
Timing [ns]	200 000	20 000	350	25	25
Particle Rate [kHz/mm ²]	100	10	250	1000	10000
Fluence [n _{eq} /cm ²]	> 10 ¹²	> 10 ¹³	10 ¹²	2x10 ¹⁵	2x10 ¹⁶
Ion. Dose [Mrad]	> 0.3	0.7	0.4	80	>500



↙ MAPS

Hybrid



See talk by M. Winter in Plenary 6, Thursday 14:00

Which technologies can be used here?
Hybrid pixels, MAPS, something new .. ?

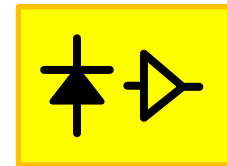
Hybrid Pixels with CMOS active sensor:

- **CMOS** (HV or HR) as a sensor (8")
- Standard (or digital) only FE chip
- Example: CCPD (HVCMOS) on FE-I4 [1]

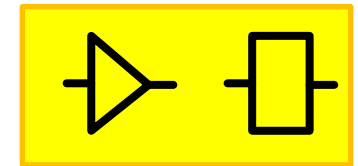
See talk by D. Münstermann
in Plenary 8, Friday 10:10

Monolithic Active Pixel Sensor

- **MAPS** [2,3] usually on non-depleted epi substrate
→ **diffusion signal**, not suited for HL-LHC
- **Charge collection by drift** → **Fully depleted MAPS (DMAPS [4])** on high resistive substrate
- Slow r/o (known from MAPS) usually does not meet LHC timing requirements

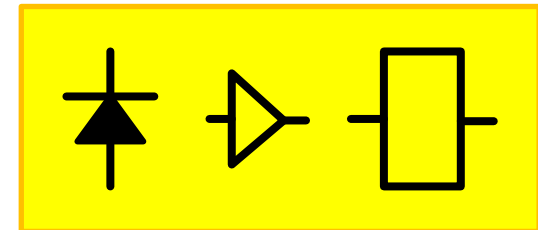


Diode +
preamp



FE chip

See poster by
T. Kishishita



Diode + Amp + Digital

→ This talk is about test results from the **sensor characterization of a DMAPS prototype**

Features

- Technology: ESPROS, 150 nm
- Deep p-type well to isolate full CMOS electronics
- 6 metal layers
- High resistive n-type bulk ($> 2 \text{ k}\Omega \text{ cm}$)
- High voltage at sensor domain possible ($\sim 20 \text{ V}$)
- Backside p-implantation
- Thinned down to $50 \mu\text{m}$

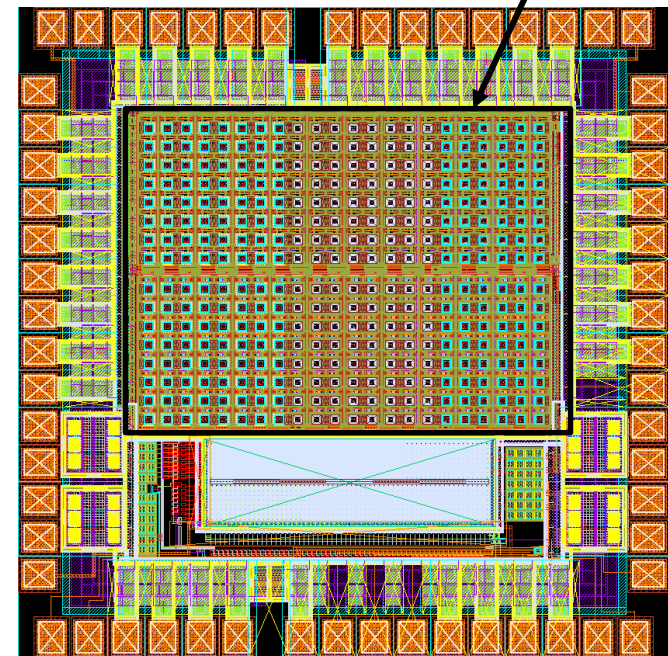
Pixel array features

- Pixel size: $40 \times 40 \mu\text{m}^2$
- Charge collecting n-implantation area $\sim 20 \times 20 \mu\text{m}^2$

Chip Dim.

$1.4 \times 1.4 \text{ mm}^2$

352 pixels

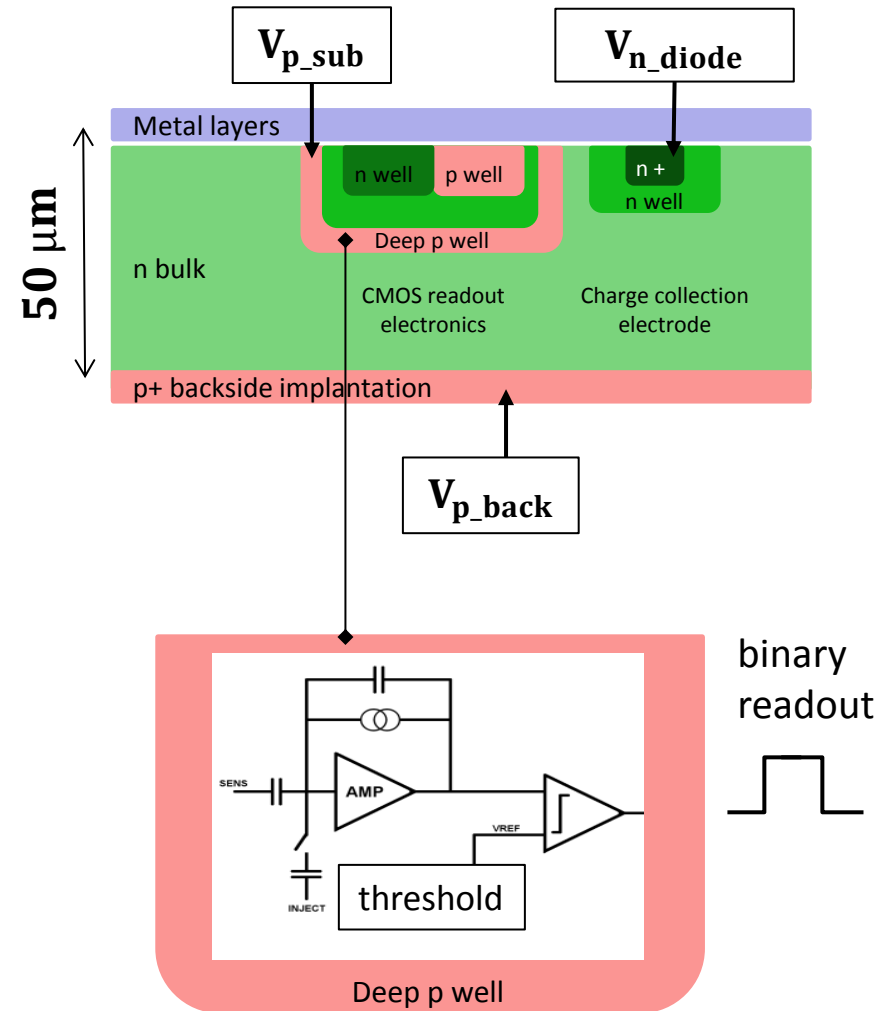


Sensor

- N-implant is charge collecting electrode
- P-backside implantation
- Full CMOS electronics inside deep p-well

Front end

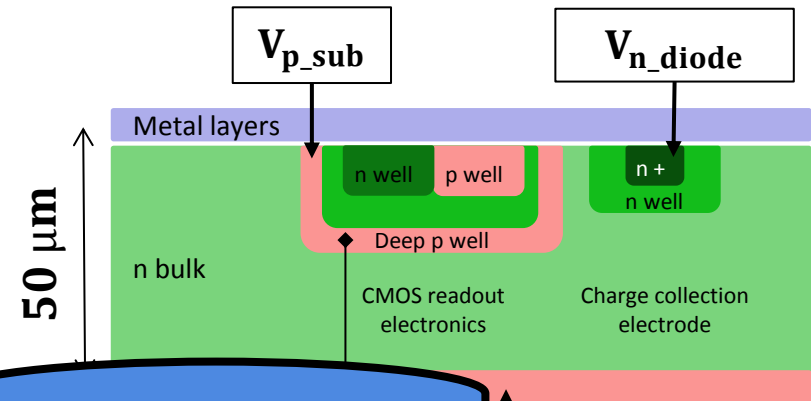
- CSA and comparator



Sensor

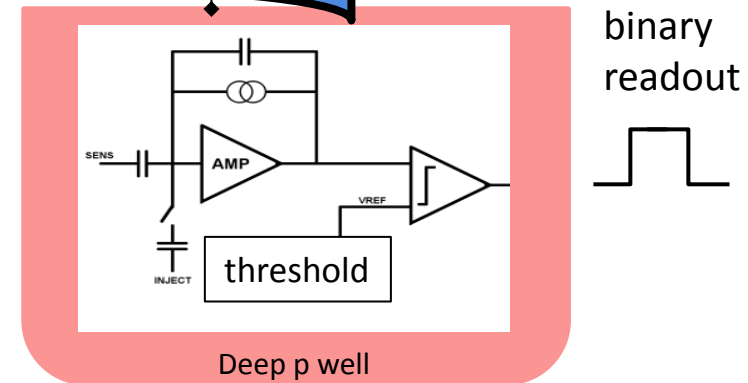
- N-implant is charge collecting electrode
- P-backside implantation
- Full CMOS electronics inside deep p-well

→ Optimization of bias voltages is needed to characterize charge collection properties !



Front end

- CSA and comparator

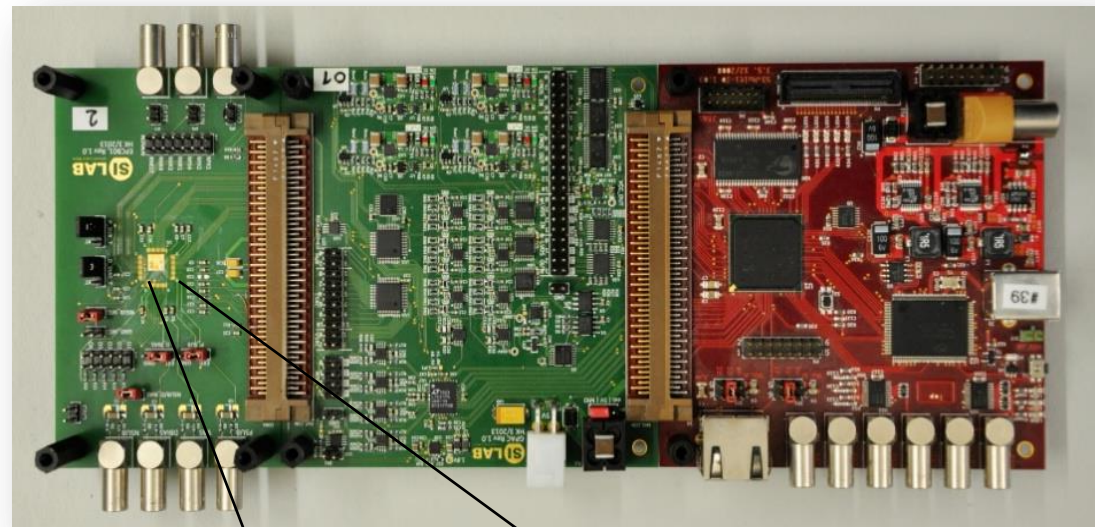


Software

- Python project EPCB01 (based on modular test framework BASIL)

Hardware

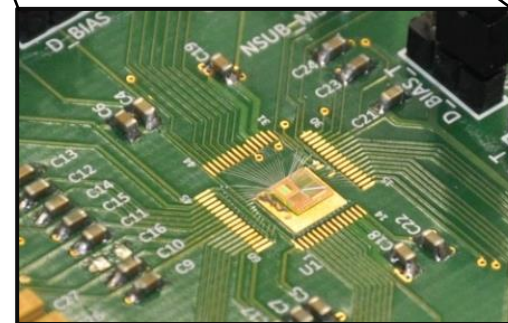
- FPGA board: S3 MultiIO-board
- General purpose adapter card (analog and digital card with voltage sources, current sources, ADC, Injection circuitry)
- Chip carrier board



Chip board

GPAC

MultiIO board



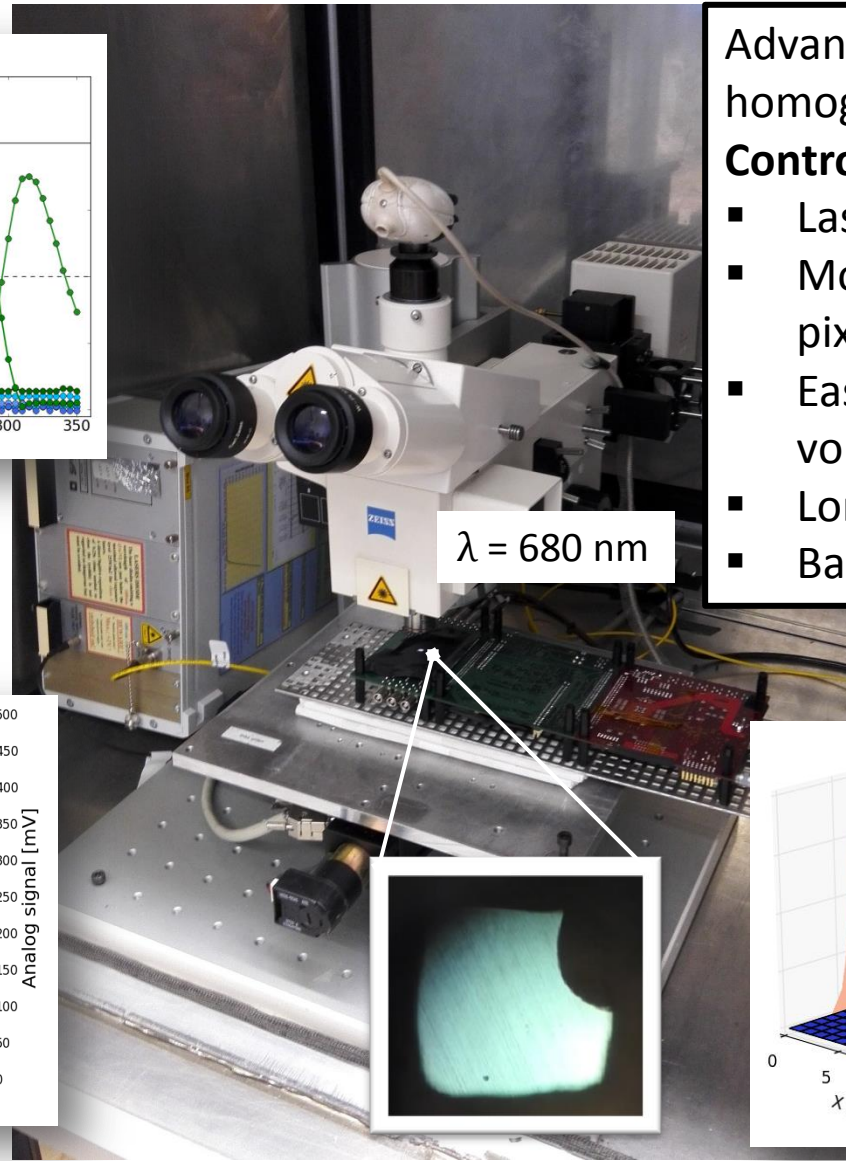
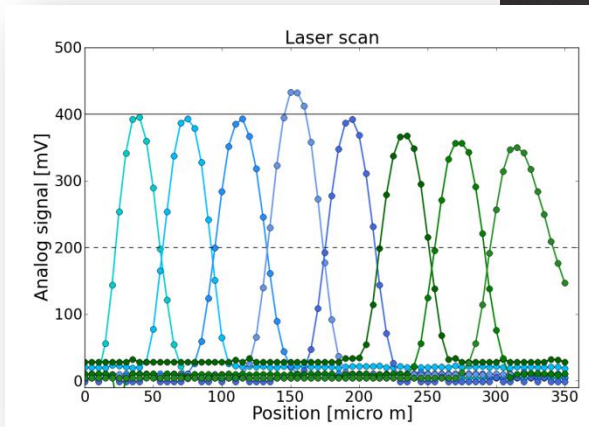
EPCB01

Laser

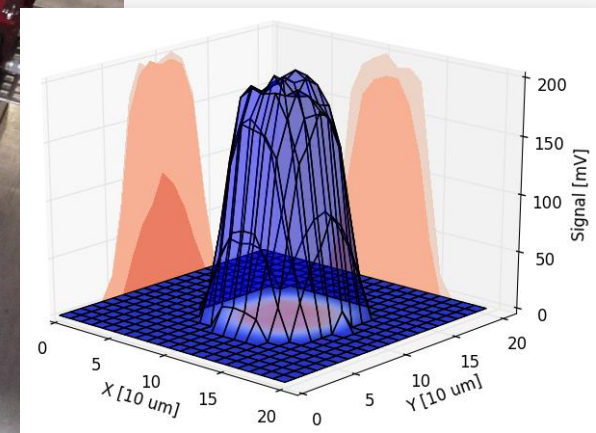
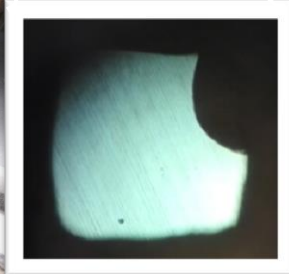
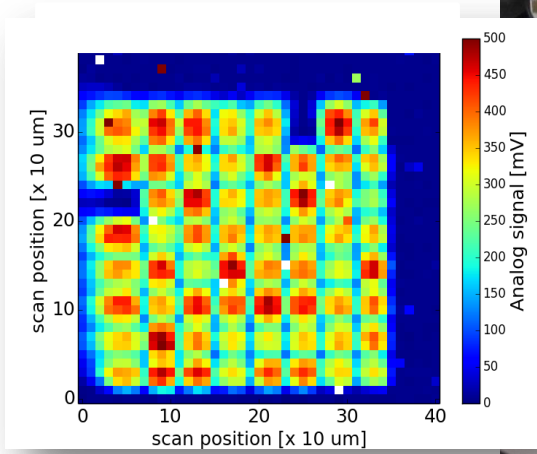
- Response homogeneity
- Bias parameter optimization

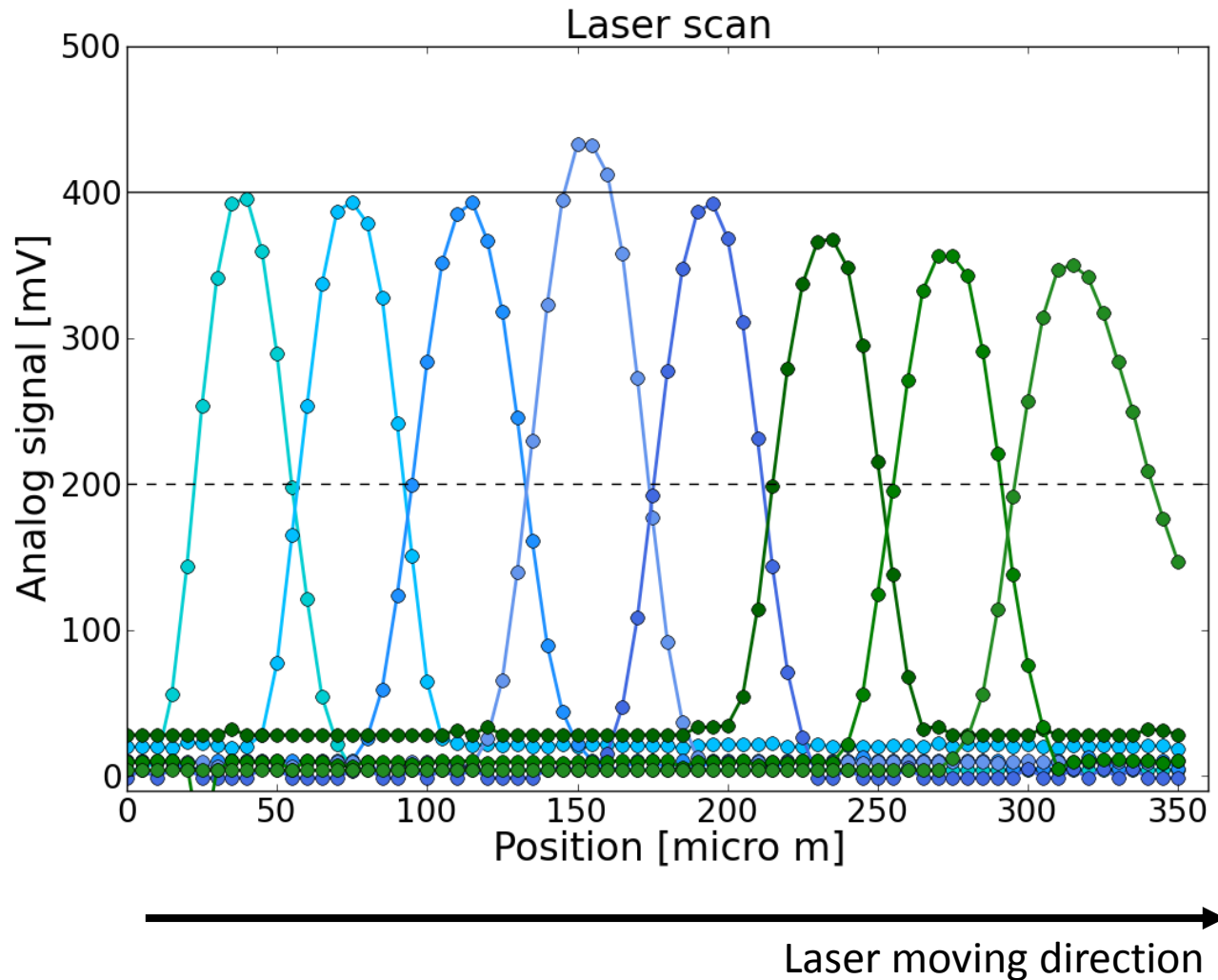
Sr 90 and electron beam

- Measure charge collected for MIP
- Cross check of laser results
- Relative CC dependence on bias parameters



- Advantages of laser system for homogeneity studies:
- Controlled environment**
- Laser spot size $\sim 3 \mu\text{m}$
 - Movable in $2 \mu\text{m}$ steps \rightarrow In pixel studies
 - Easy repetition \rightarrow Fast voltage scans
 - Long term stability $\pm 100 \text{ e}$
 - Backside illumination



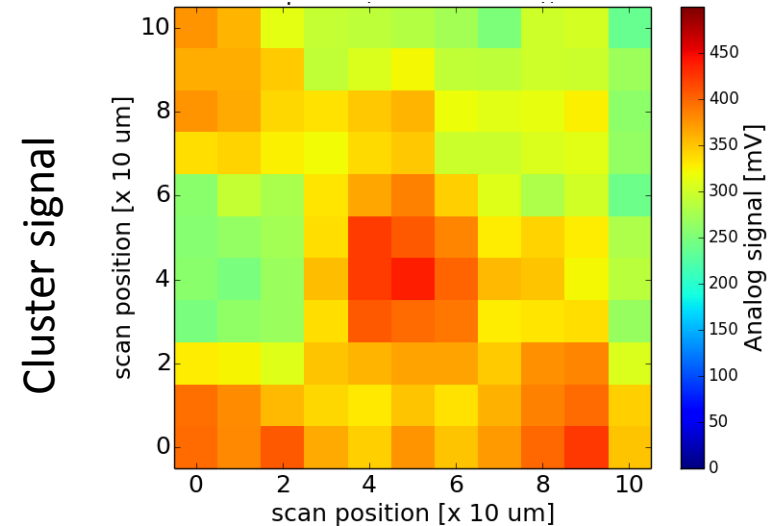
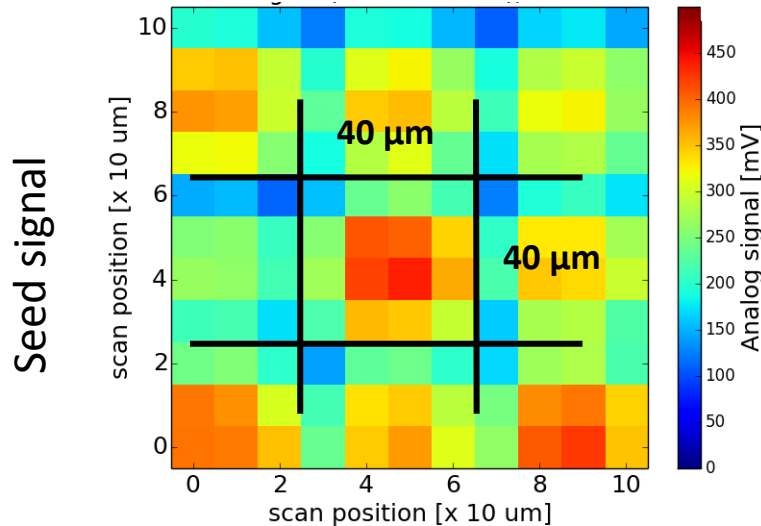
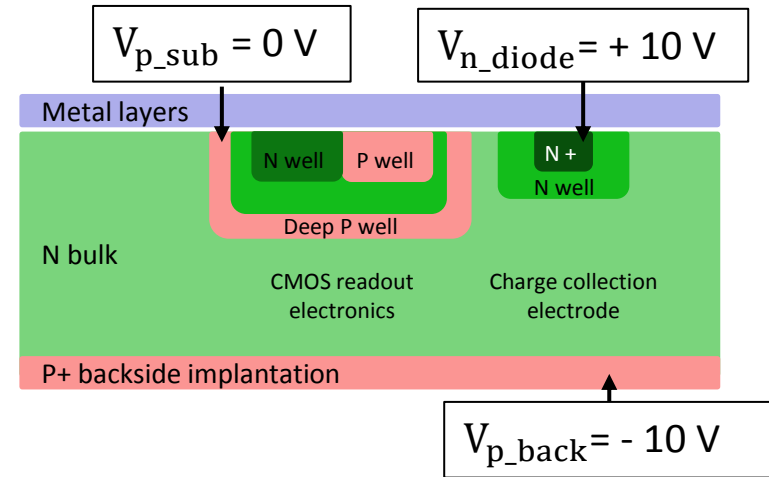


Laser scan over nine pixels

Scan settings

- Step size: 10 μm , 10 x 10 steps

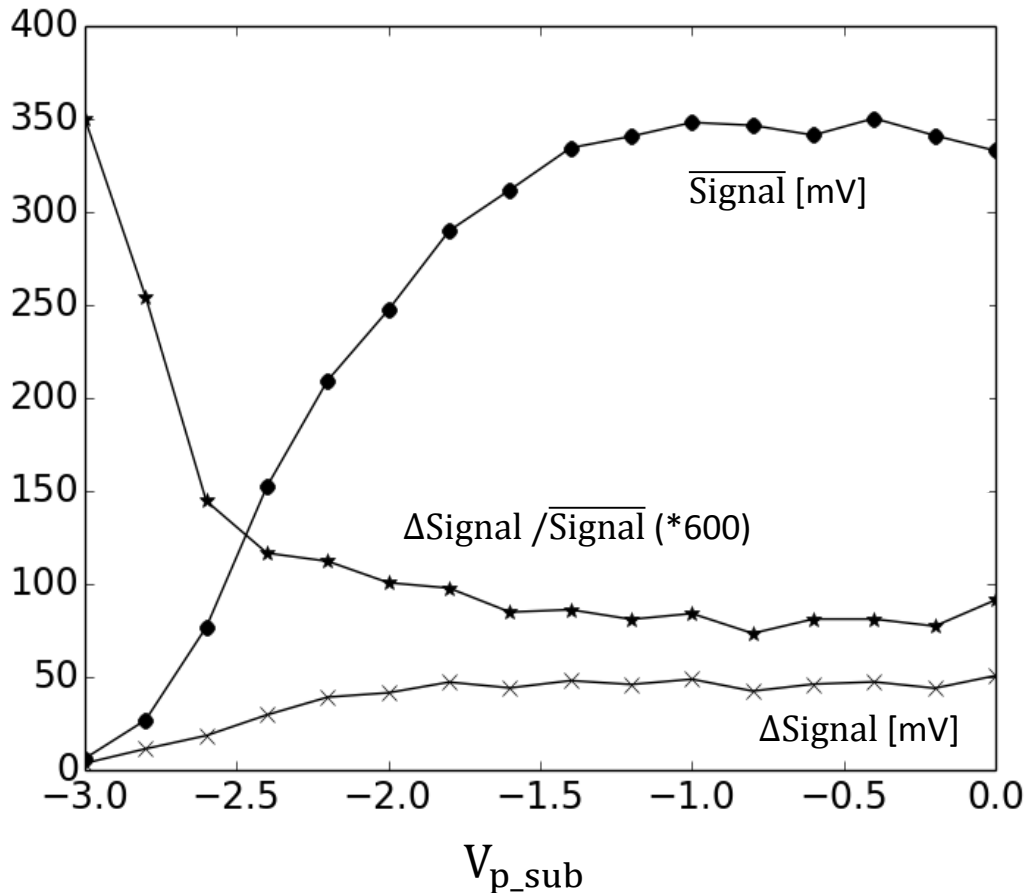
Remark: No fixed pattern noise correction (worst case)



→ Figure of merit: $\frac{\Delta\text{Signal}}{\overline{\text{Signal}}}$

Parameter scan of V_{p_sub} over nine pixels

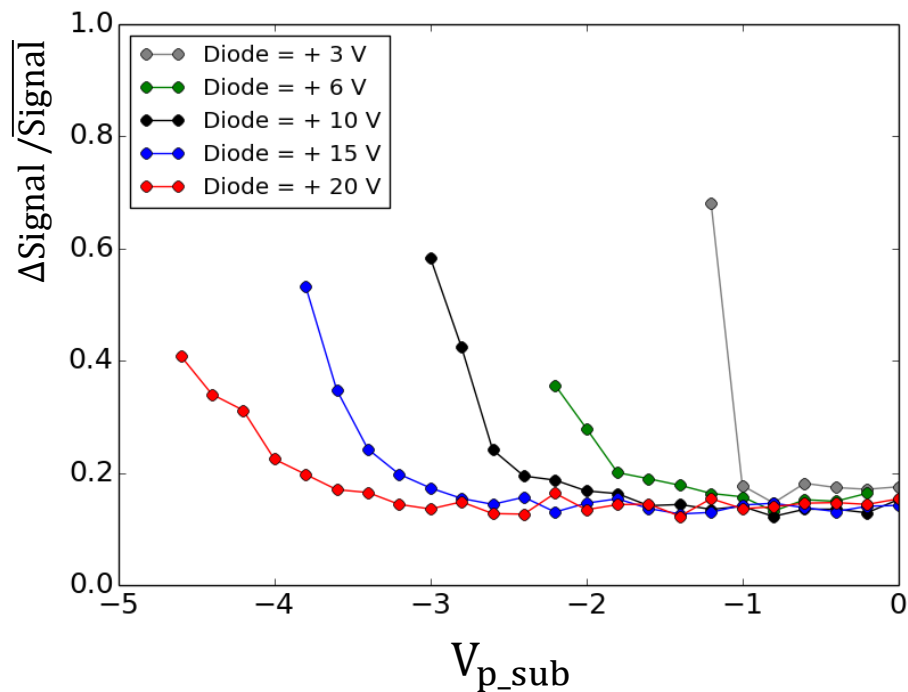
Δ Signal and $\overline{\text{Signal}}$ versus V_{p_sub}



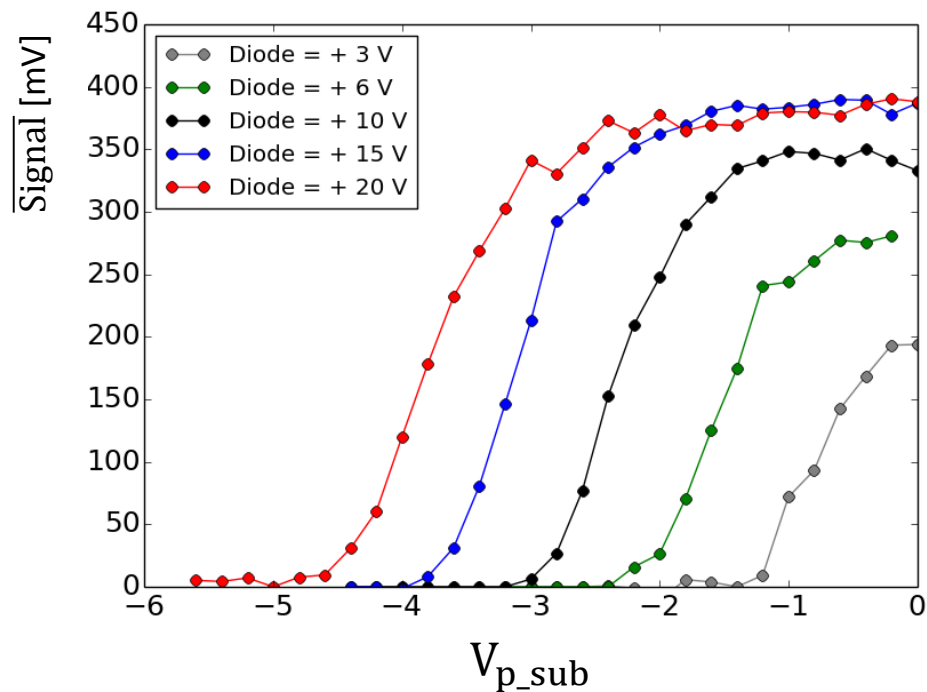
- Best homogeneity
 Δ Signal / $\overline{\text{Signal}}$ = 15 %
for V_{p_sub} from 0 to - 1.5 V
- No charge collection
below $V_{p_sub} = - 3$ V

Parameter scan of V_{p_sub} and V_{n_diode} over nine pixels

Inhomogeneity



Amplitude

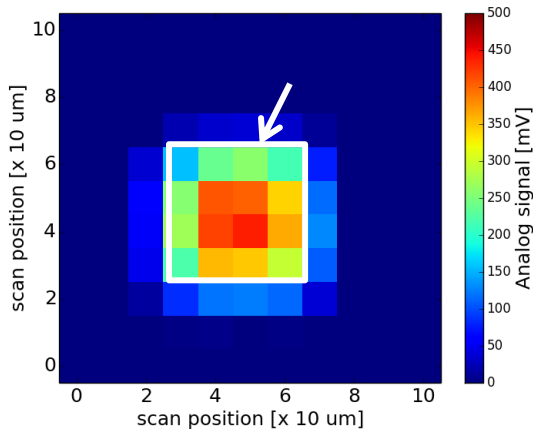


→ Depleted sensor at $V_{n_diode} > +15 \text{ V}$

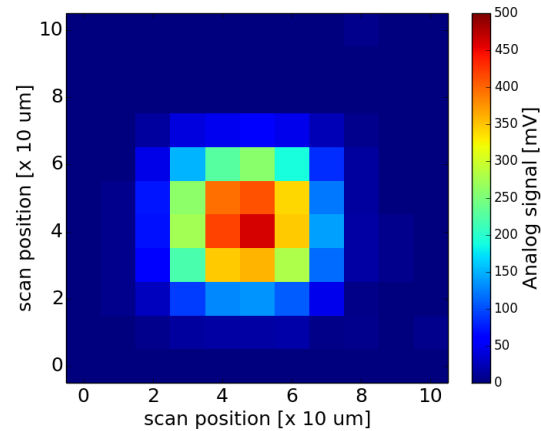
→ With increasing V_{n_diode} , the range for V_{p_sub} with signal increases

Signal of central pixel for different V_{p_sub}

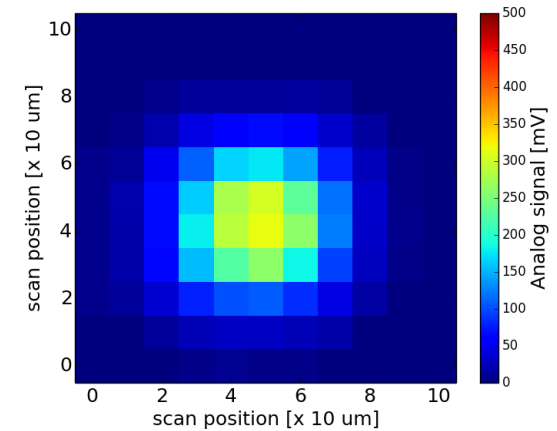
$V_{p_sub} = 0 \text{ V}$



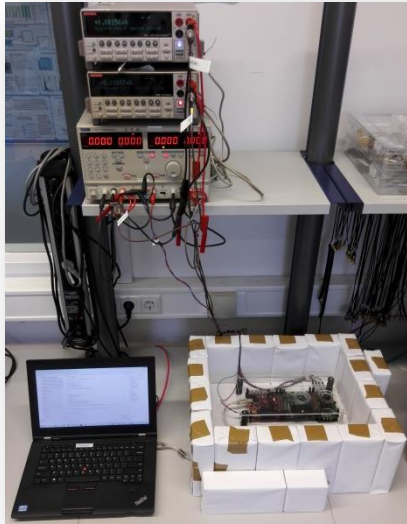
$V_{p_sub} = - 1 \text{ V}$



$V_{p_sub} = - 2 \text{ V}$



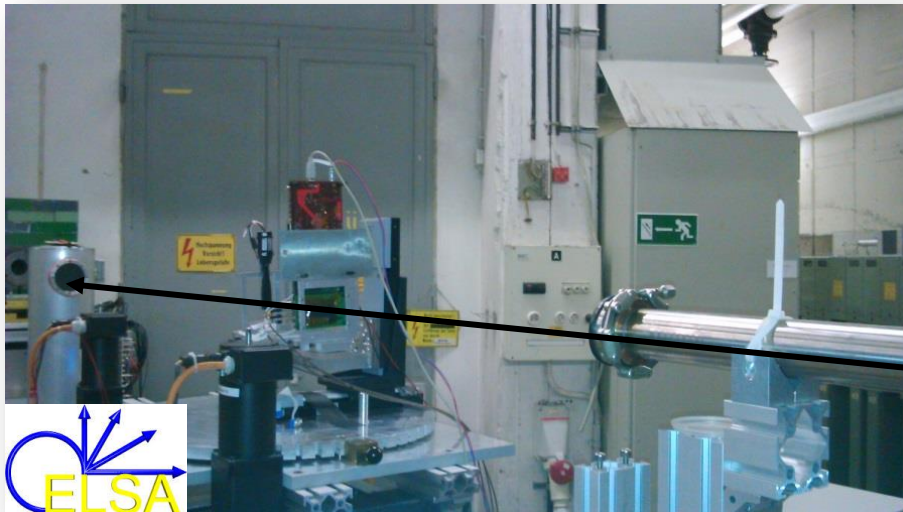
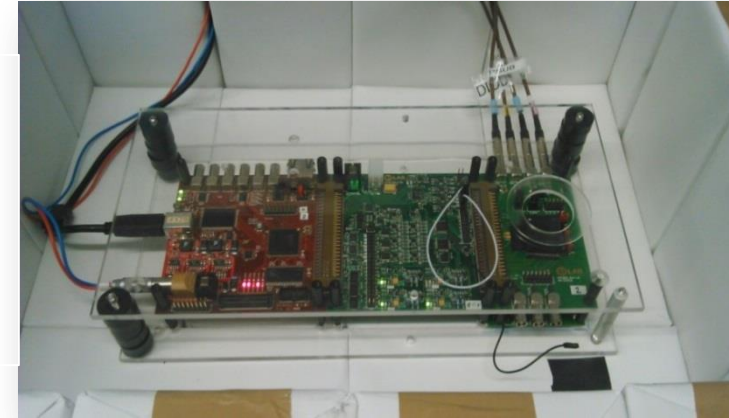
→ Lowering V_{p_sub} the charge spread is larger and the seed signal is decreased



← Source setup →

Sr90

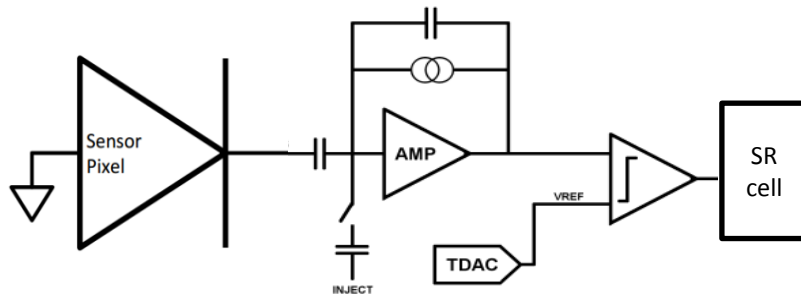
Expect $Q_{\text{Sr90}}(\text{MPV}) = 3.8 \text{ ke}$
in $50 \mu\text{m}$ silicon



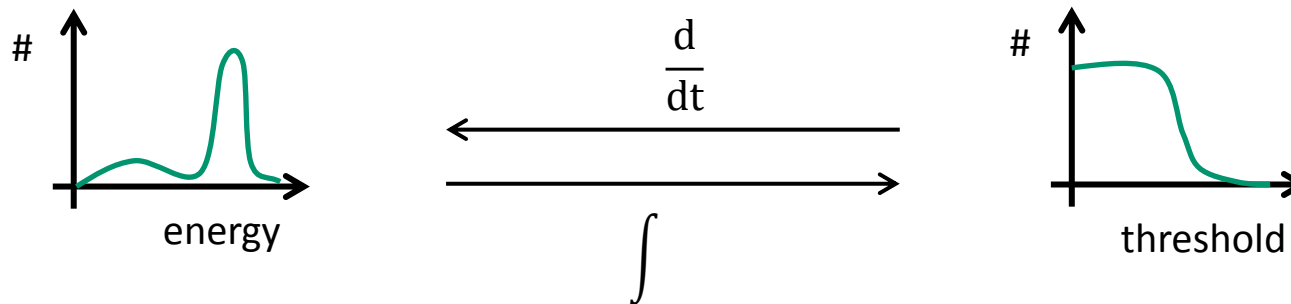
← Electron beam setup

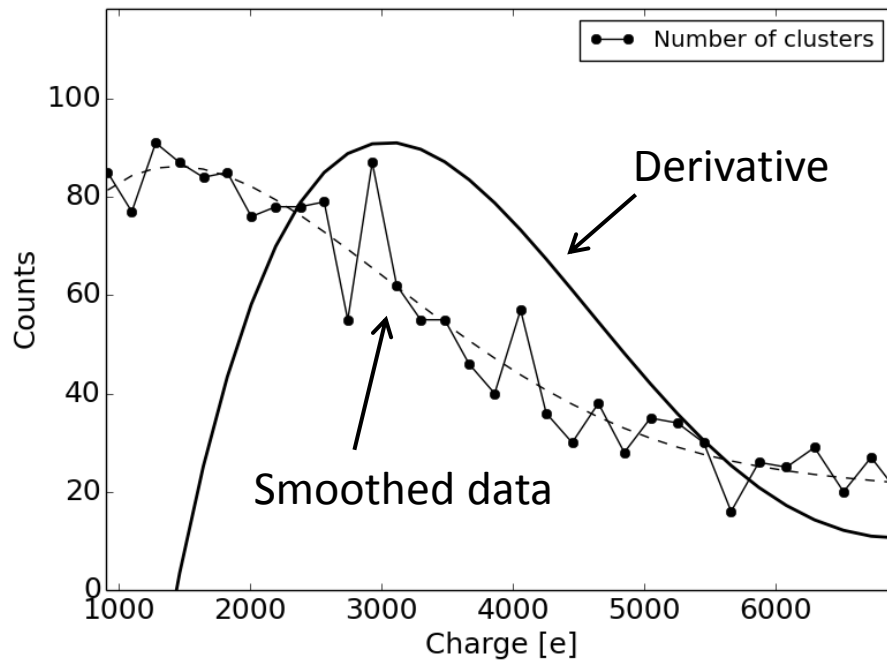
3.2 GeV electron beam

Expect $Q_{\text{MIP}}(\text{MPV}) = 3.3 \text{ ke}$
in $50 \mu\text{m}$ silicon



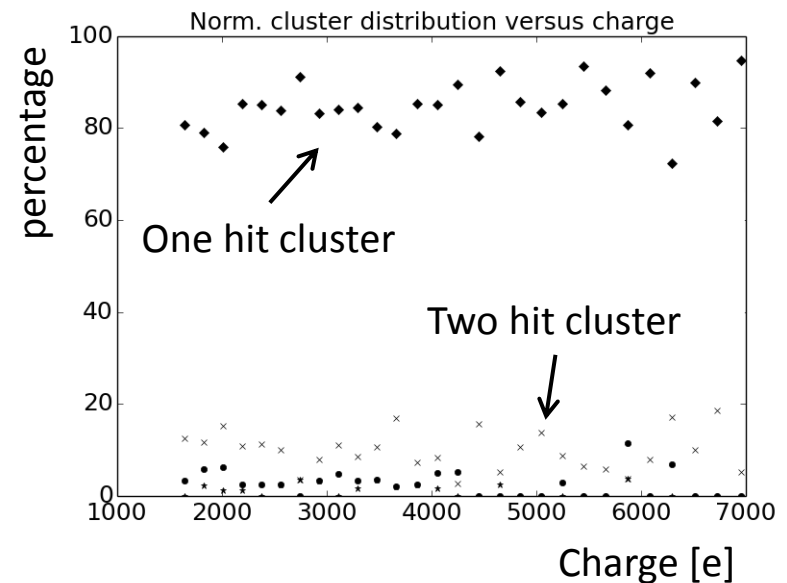
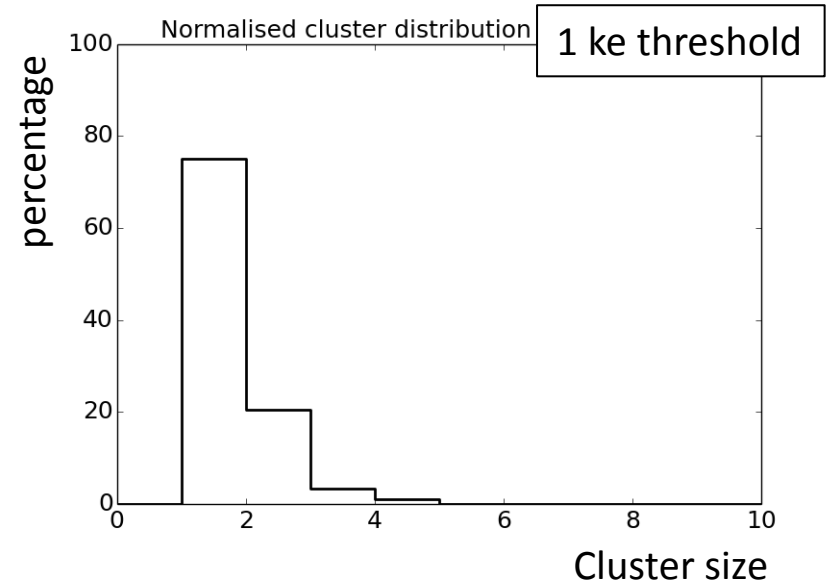
- Count cluster versus threshold voltage
- No trigger
- Analysis with 36 pixels
- Calibrate x-axis with internal injection, global calibration with cut at 7 ke



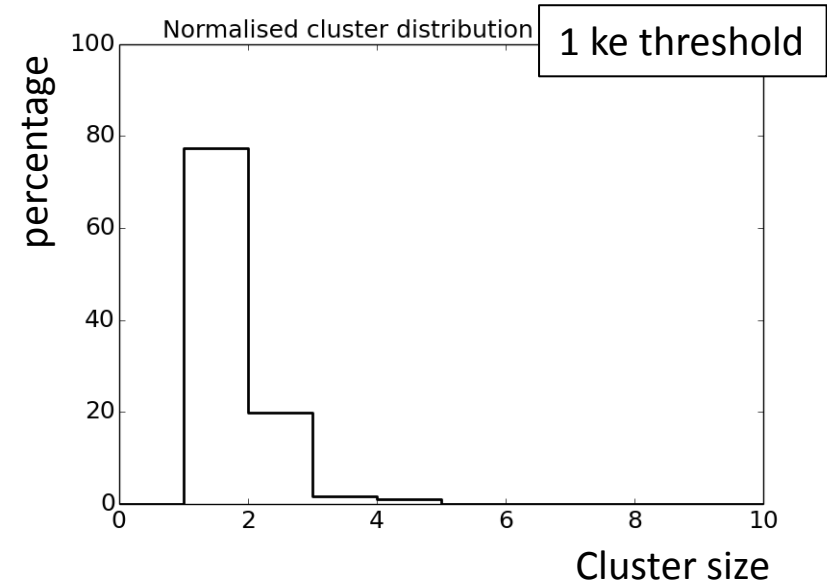
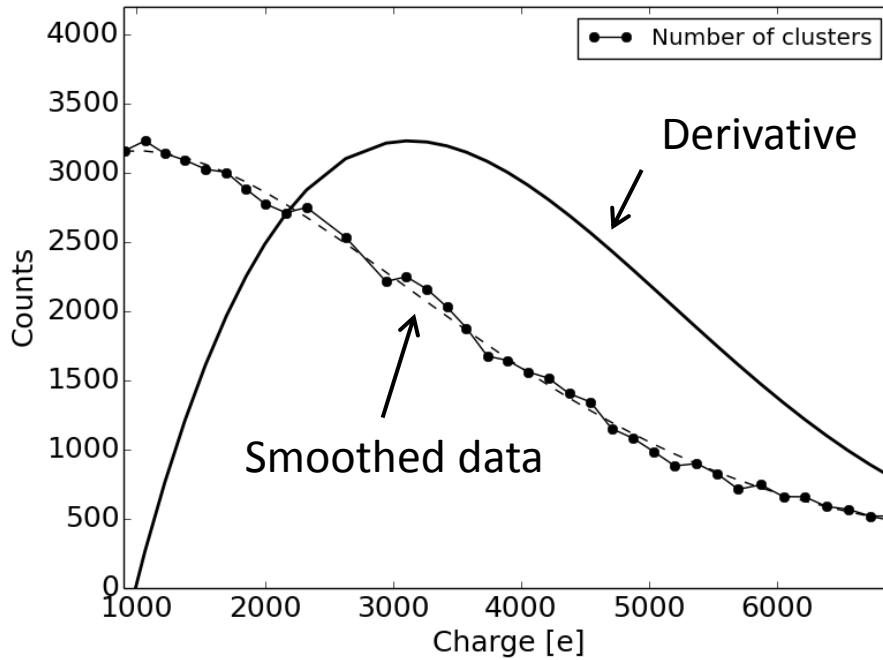


Reconstructed charge:

- 3.1 ke
- 15% less than expected (3.8 ke)

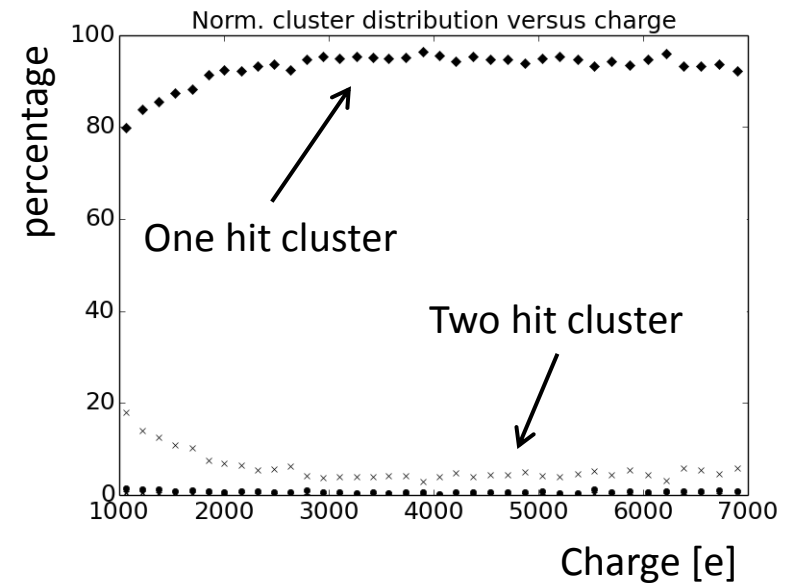


Results of 3.2 GeV electron beam (ELSA) measurement

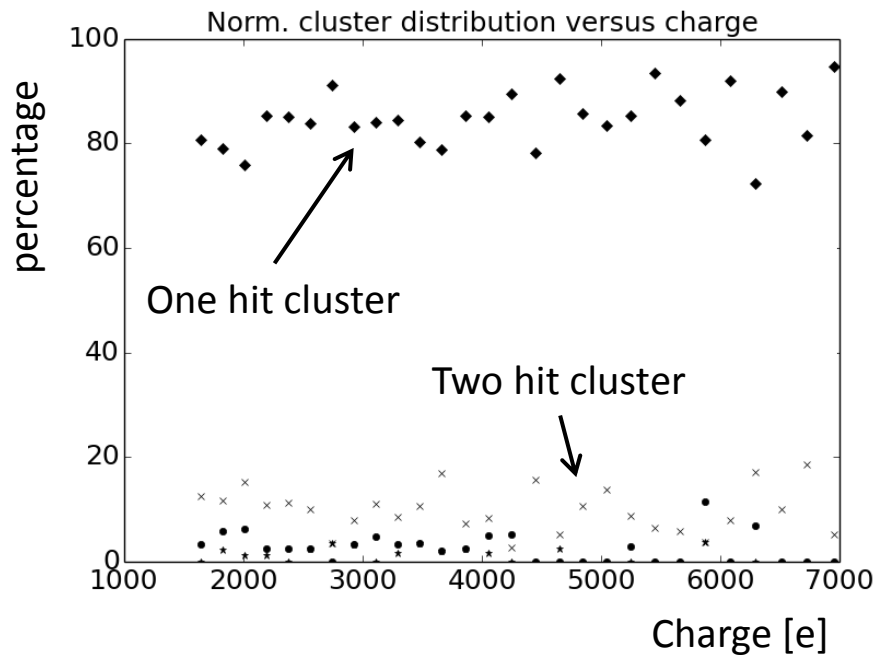


Reconstructed charge:

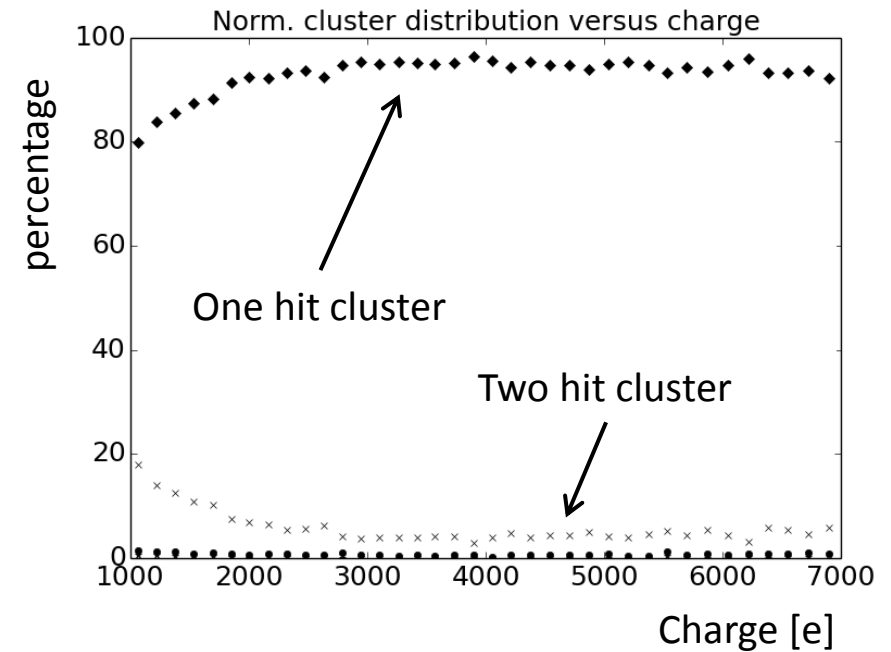
- 3.1 ke
- 7% less than expected (3.3 ke)
- At highest threshold rate is not zero (delta electrons)



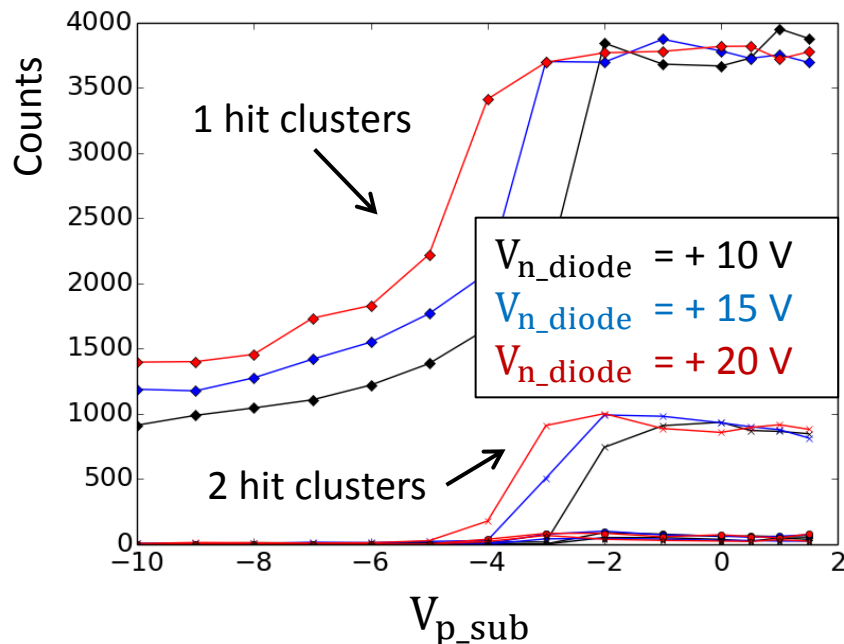
Sr 90 collimated



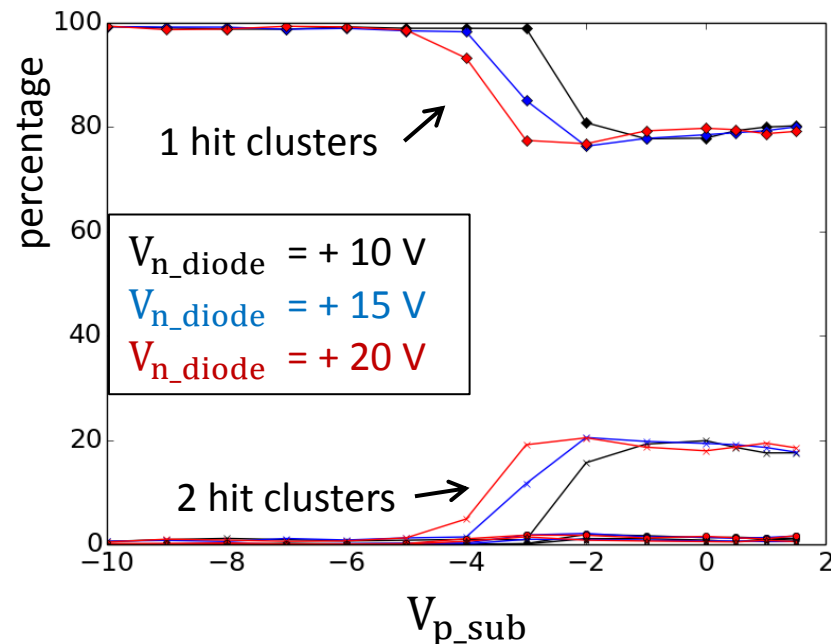
3.2 GeV electron beam



Absolute cluster distribution



Normalised cluster distribution



- At low V_{p_sub} the one hit cluster count rate drops to $\sim \frac{1}{4}$
- At low V_{p_sub} the cluster distribution changes from 80 % to $\sim 100\%$ one hit clusters
- Count rate decrease starts later with higher V_{n_diode}

Observations:

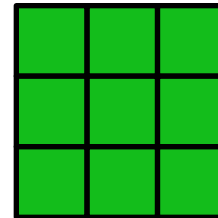
- Count rate decreases to $\frac{1}{4}$ with decreasing V_{p_sub}
- Only one hit clusters at lower V_{p_sub}
- Count rate decrease starts ‚later‘ with increasing V_{n_diode}

Fraction of count rate loss corresponds roughly to the fraction of n-implantation area to total area ($\sim \frac{1}{4}$)

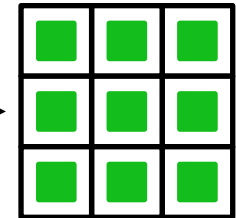
→ The inefficient regions are likely to be below the psubstrate.

Charge is partially lost.

$P_{sub} = 0 \text{ V}$



$P_{sub} = -6 \text{ V}$



 Efficient region

 Inefficient region

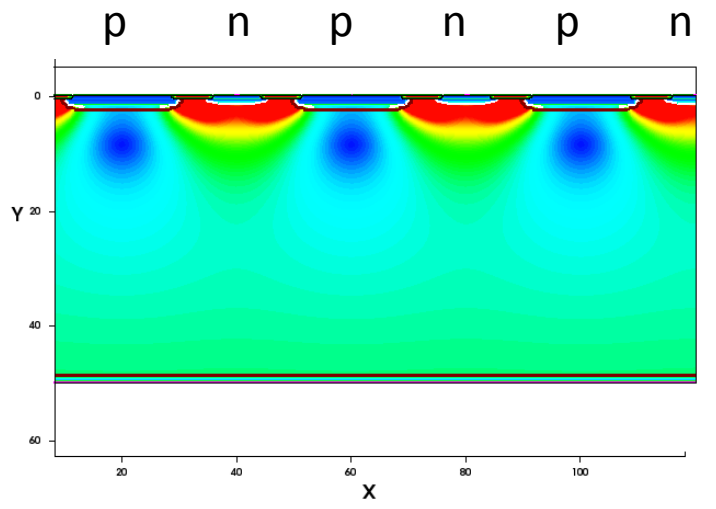
→ Inefficient regions start to built up with decreasing V_{p_sub}

Possible reasons for charge loss:

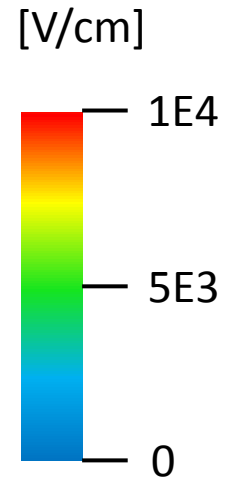
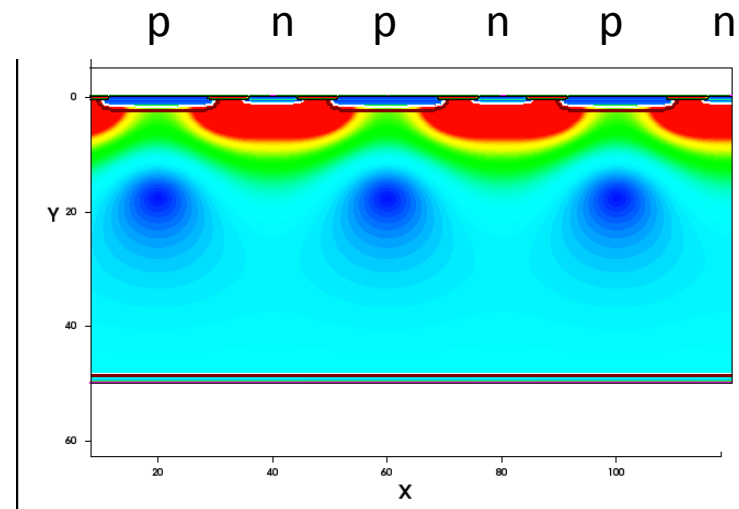
- Trapping → Unlikely in unirradiated sensor
- Slower charge collection → Finite integration time of CSA of $\sim 10 \text{ ns}$
- Competing electrode → Unlikely as psub implant is not attractive to electrons

→ **Next step:** TCAD Simulation (in progress)

$$V_{p_sub} = 0 \text{ V}$$



$$V_{p_sub} = -10 \text{ V}$$



- Minimum of electric field shifts with lower V_{p_sub} towards backside
- Electric field is close to zero at the backside for $V_{p_sub} = -10 \text{ V}$

→ Effect on charge collection time is work in progress

Sensor characterization of the first DMAPS prototype in ESPROS technology has been performed

- Homogeneity studies and bias parameter optimization with laser system
- Sr 90 and electron beam used for measurement of charge collection
- Biasing of the sensor is partially understood

Radiation hardness studies are work in progress

- Xray irradiation studies show transistors are radiation tolerant up to 50 Mrad TID
- Neutron irradiated samples arrived recently (up to $5E+14$ neq)

New version with improved electronics submitted in Q2 2014, expected to arrive in Q4 2014

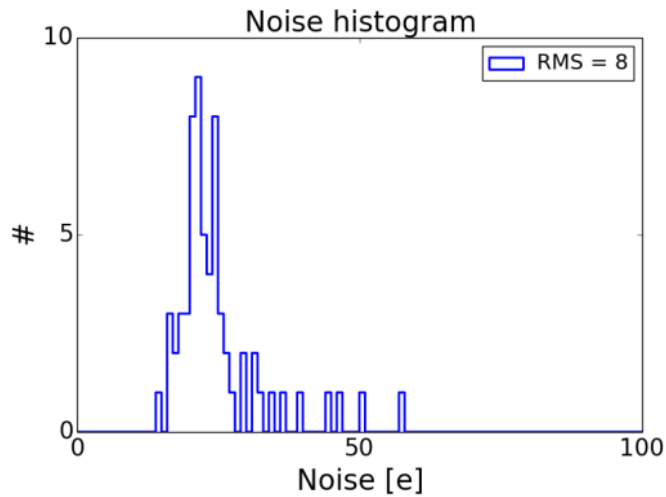
Thank you

- [1] I. Peric and C Takacs, *Large monolithic particle pixel detectors in high-voltage CMOS technology. Nucl. Instrum. Methods A 624 (2010), 504-508*
- [2] S. Senyukov et al., *Charged particle detection performances of CMOS pixel sensor produced in a 0.18 μm process with a high resistivity epitaxial layer. Nucl. Instrum. Methods A 730 (2013), 115-118.*
- [3] R. Turchetta et al., *A monolithic active pixel sensor for charged particle tracking and imaging using standard VLSI CMOS technology. Nucl. Instrum. Methods A 458 (2001), 677-689.*
- [4] M. Havranek et al., *DMAPS: a fully depleted monolithic active pixel sensor – analog performance, [2014 JINST 1 P0514](#).*

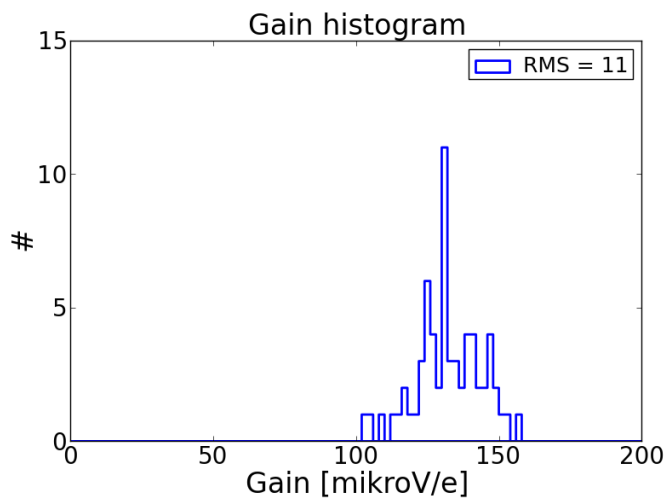
Related papers:

- [5] ATLAS Collaboration. *ATLAS pixel detector electronics and sensors. JINST 3 (2008)*
- [6] Y. Mikami, *Monolithic Active Pixel Sensors (MAPS) in a Quadruple Well Technology for Nearly 100% Fill Factor and Full CMOS Pixels. Sensors 8 (2008), 5336-5351.*
- [7] H. Bichsel, *Straggling in thin silicon detectors*, Reviews of Modern Physics, Vol. 60, No. 3, July 1988.
- [8] Pohl et al., *A method for precise single pixel charge reconstruction for counting pixel detectors*, To be submitted

Backup



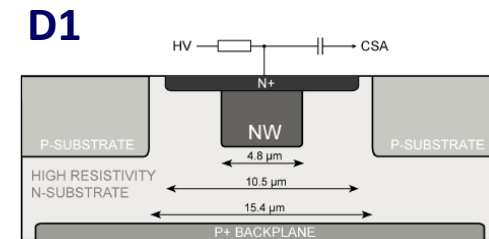
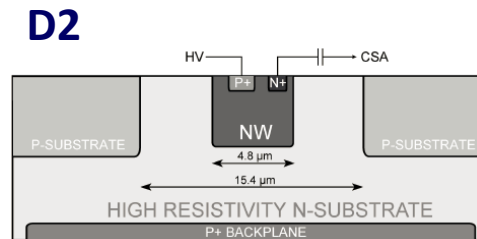
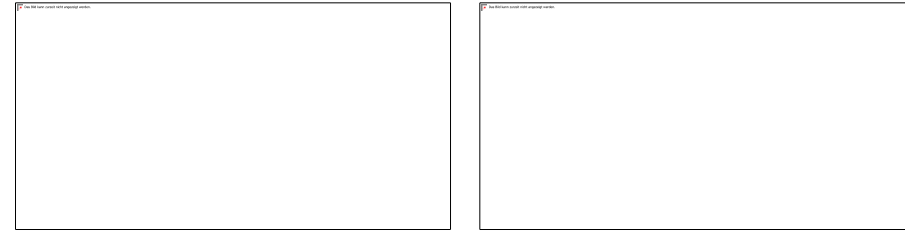
→ Noise ~ 25 e



→ Gain ~ 125 $\mu\text{V}/\text{e}$

■ EPCB02 is evolution of EPCB01

- cascode amplifier with high open loop gain
- added 2 fF feedback capacitor -> uniformity
- both charge collection electrodes are custom
- input transistors are larger
- each pixel has analog output
(multiplexed → unity gain buffer → IO pad)



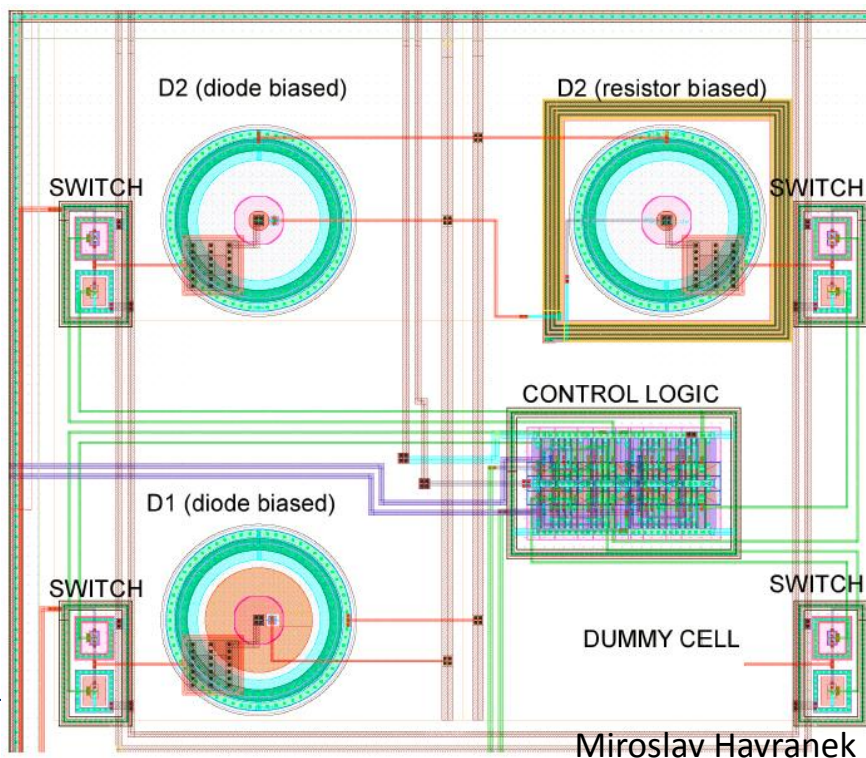
■ EPCB02 will allow to study independently:

- V1 vs V2 – effects of biasing (RB/DB)
- V1 vs V4 – cont. vs switched FE
- V4 vs V5 – different sensor geom.
- V4 vs V6 – different transistor size

Variant	Sensor	Biasing	FE-type	M1 dim.
V1	D2	DB	CONT	1μ/300n
V2	D2	RB	CONT	1μ/300n
V3	D1	DC	CONT	1μ/300n
V4	D2	DB	SW	1μ/300n
V5	D1	DB	SW	1μ/300n
V6	D2	DB	SW	2μ/150n

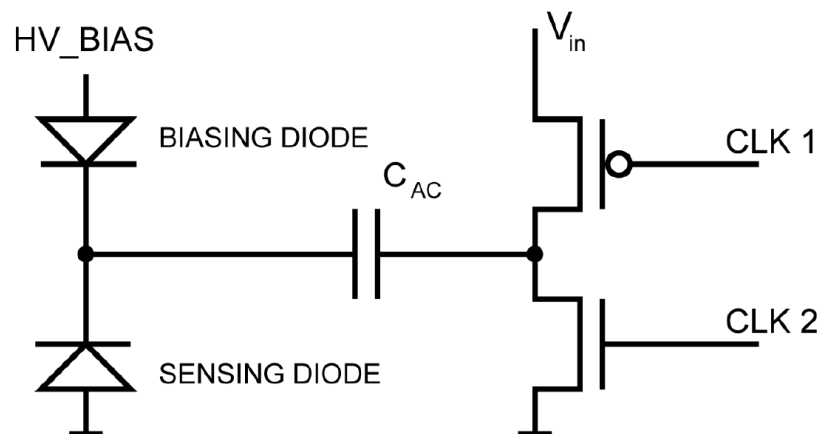
Circuit for capacitance measurement in EPCB02

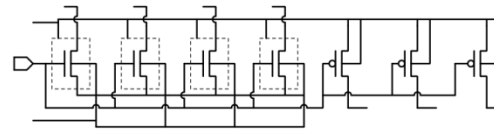
- Various layouts and biasing circuits can influence sensor capacitance
- C_d is important parameter for rise-time and noise optimization
- Charge-pump based circuit for capacitance measurement of DMAPS sensor



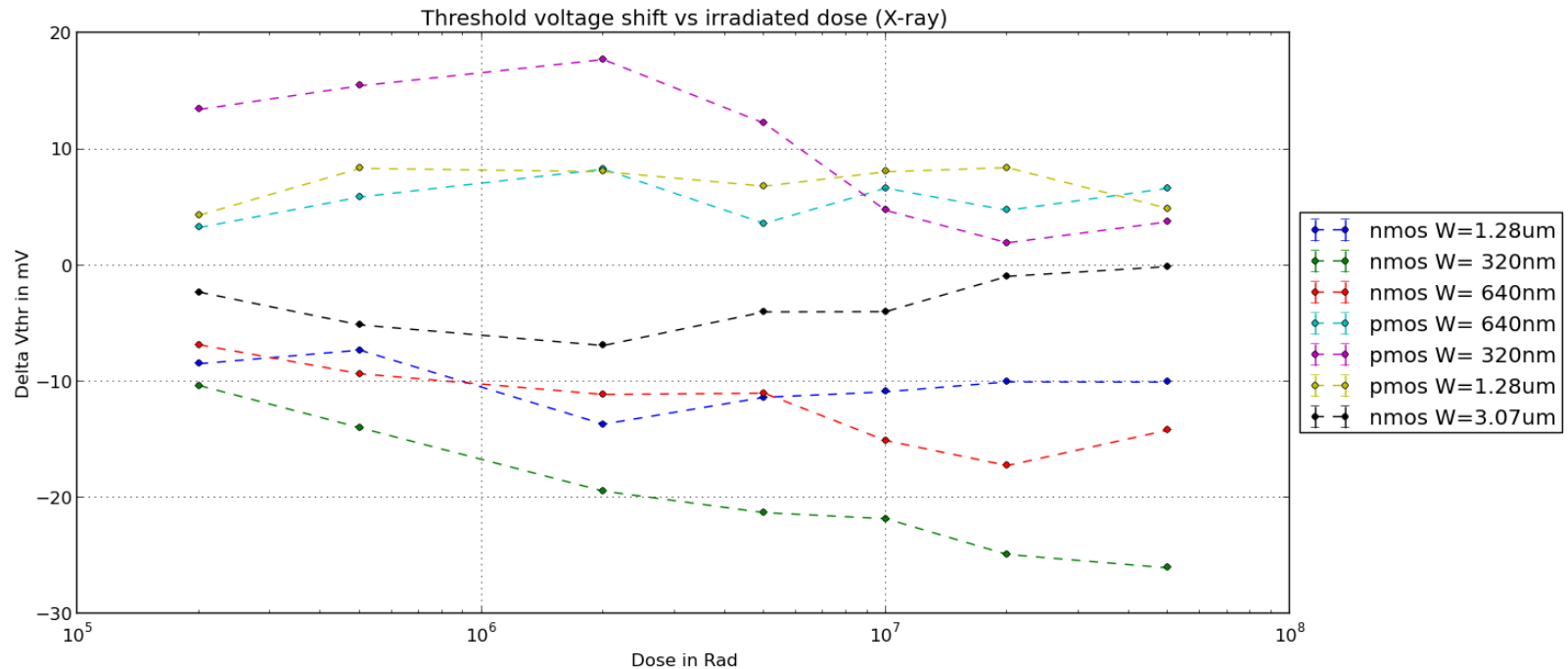
- **Two transistor charge pump:**

- CV measurement => depl. voltage.
- compare different geometries





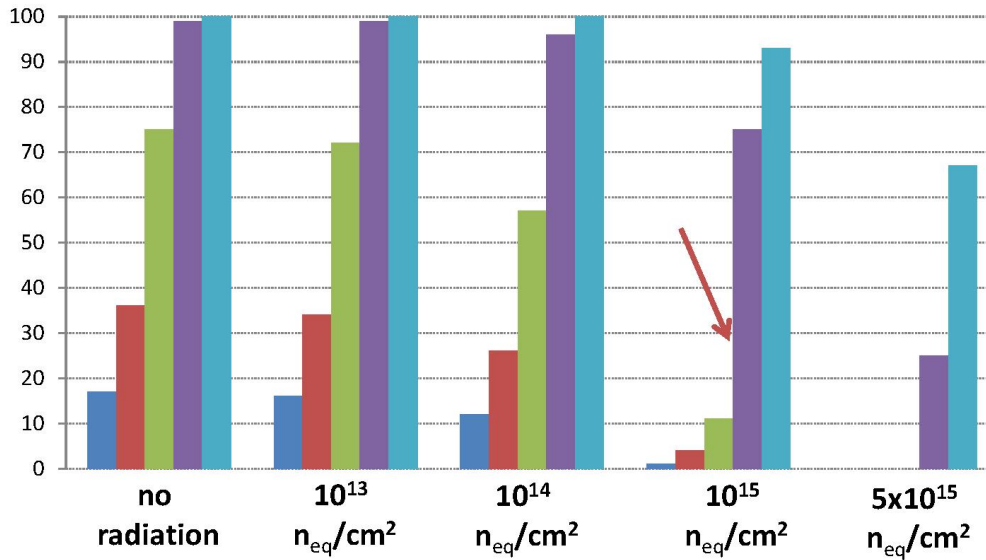
- Irradiation with X-Ray (60 keV) and Vanadium filter (peak $\sim 5\text{keV}$) up to 50 Mrad



- **Less than 25 mV shift after 50 Mrad of Total Ionizing Dose (TID)**
- Enclosed layout has lower threshold shift and recovers at 50 MRad

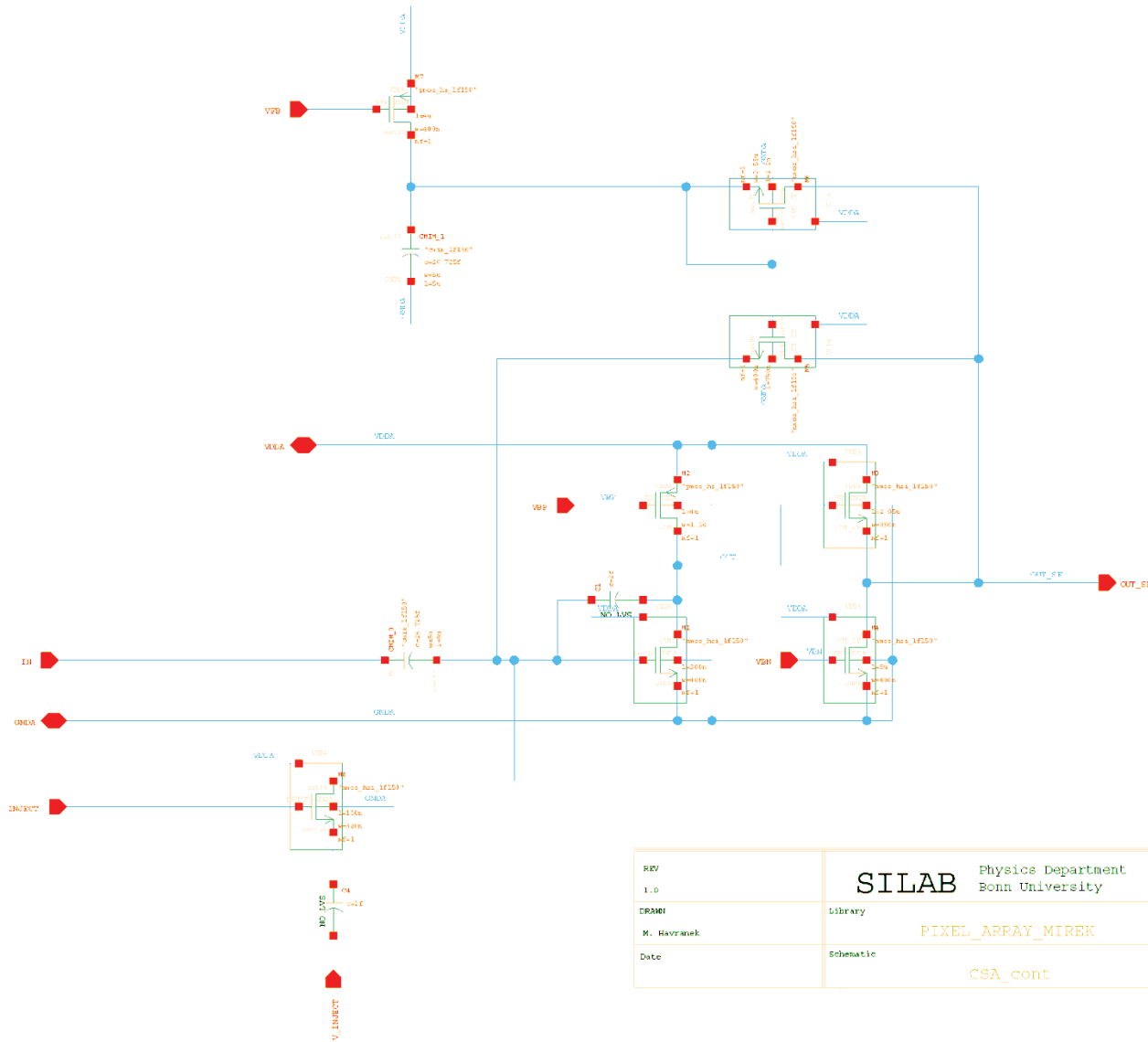
TCAD simulation of CC Time in dependence of fill factor and bias voltage and resistivity

% of collected charge in first 10ns



	substrate resistivity [Ohm cm]	Bias [V]	Fill Factor [%]
Blue	10	1	15
Red	10	20	15
Green	2k	1	15
Purple	2k	20	15
Cyan	2k	20	75

Transistor level schematic of CSA



REV	1.0	SILAB	Physics Department Bonn University
CPANEL		Library	
M. Havranek		PIXEL_ARRAY_MIREK	
Date		Schematic	
			CSA_cont

Literature:

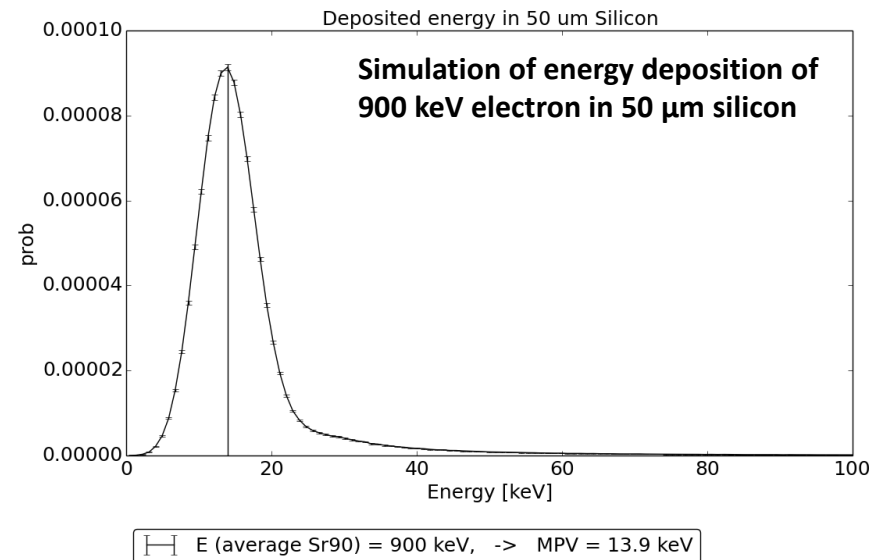
- The most probable value of energy deposition (MPV) for a **MIP** in 50 μm silicon is:
MPV = 12 keV [1]
- $\rightarrow Q_{\text{MIP}}(\text{MPV}) = 3.34 \text{ ke}$

Simulation of MPV for Sr90 source with pyPENELOPE for average energy of emitted energy spectrum Sr90 (900 keV) [2]:

MPV = 13.9 keV

- $Q_{\text{Sr90}}(\text{MPV}) = 3.86 \text{ ke}$

- $Q_{\text{generated, MIP}}(\text{MPV}) < Q_{\text{generated, Sr90}}(\text{MPV})$



[1] Bichsel, Straggling in thin silicon detectors, Reviews of Modern Physics 60 (1988) 663-699 (page: 685 and 691)

[2] Sempau, Monte Carlo simulation of electron beams from an accelerator head using PENELOPE, Phys. Med. Biol. 46 (2001) 1163-1186

Inner layer

1. **Low power**
2. **Low material**
3. Occupancy
4. Resolution



hybrid pixels (65nm? + sensor?)

Outer layer

1. **Low cost**
2. **Low power**
3. Low material
4. Resolution



low cost hybrid pixels or monolithic?

Method:

- Setup:

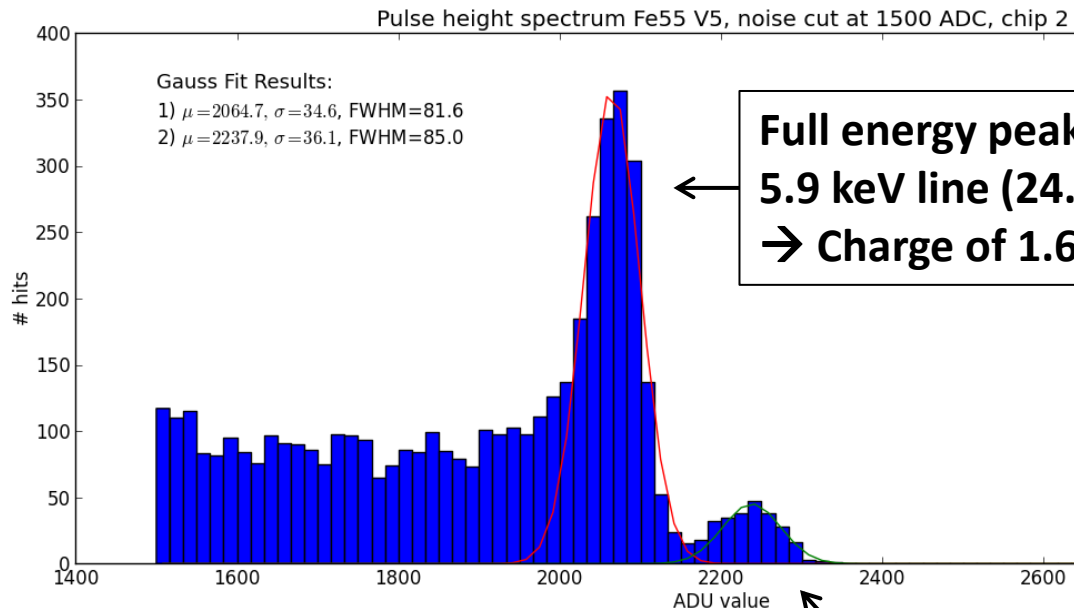
Illumination with Fe55 from the backside

- Data taking for 10 hours
- Histogram of signal
- Cut at 1500 to reduce noise

→ Both γ lines from Fe55 clearly visible in spectrum

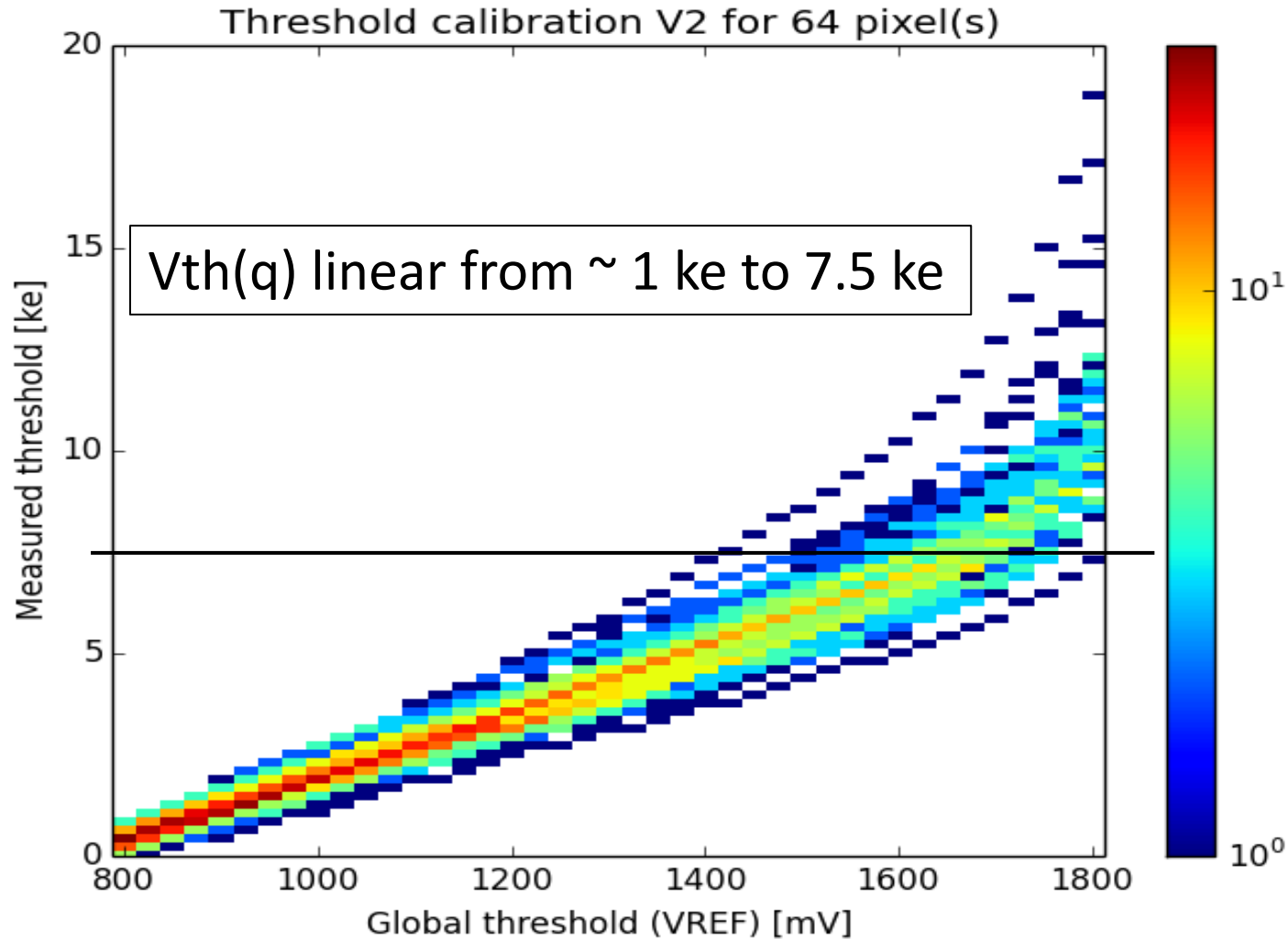
→ Sigma \sim 36 ADU (FWHM \sim 85 ADU)

→ \sim 1.1 ADU / electron

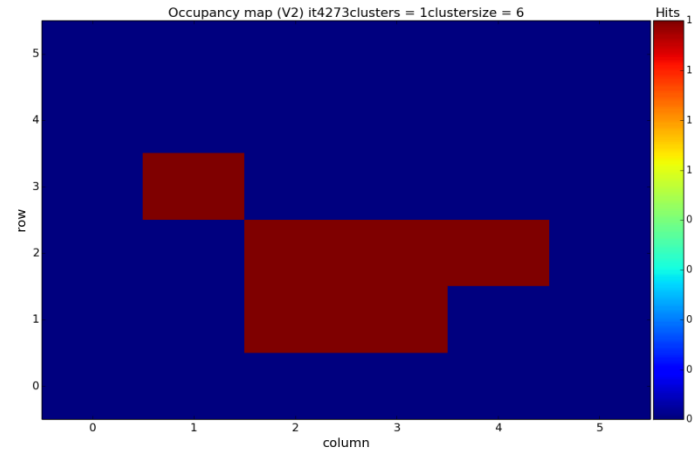
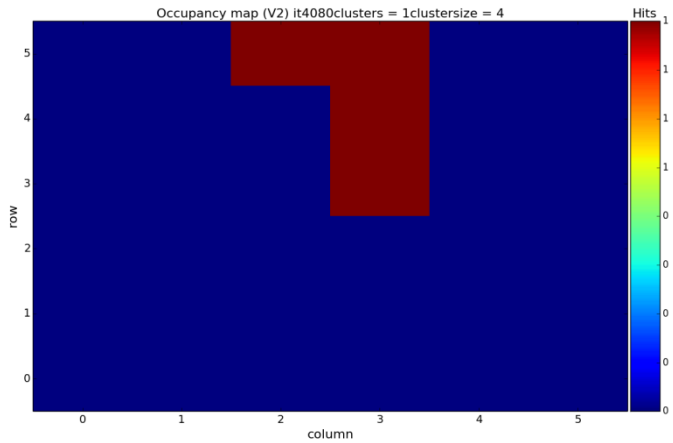
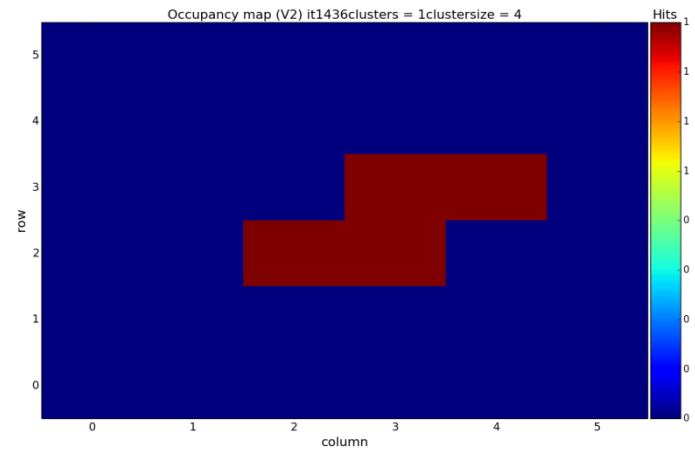
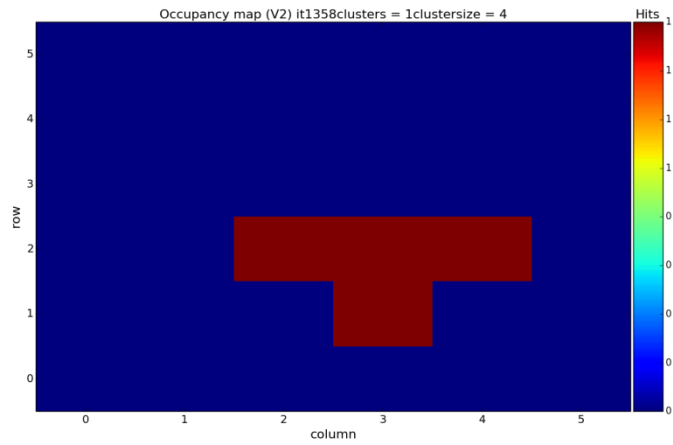


Full energy peak:
5.9 keV line (24.4%)
→ Charge of 1.6 ke

Full energy peak:
6.5 keV line (2,85%)
→ Charge of 1.8 ke

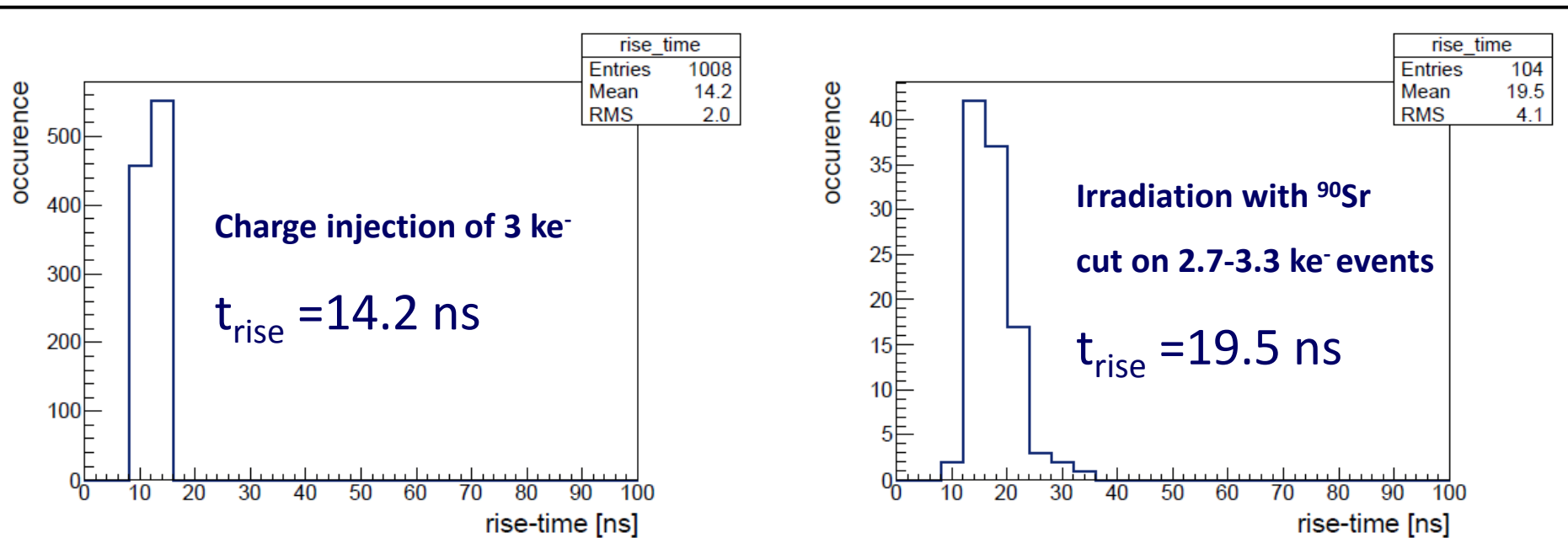


Exotic cluster examples



Signal rise-time

- Rise-time => 25% - 75% signal amplitude
- Upper limit on charge collection time



- Fast charge collection
- 50 % of the signal charge is collected for less than 19.5 ns

Effect of charge sharing and multiple scattering on the reconstructed charge

Multiple scattering → Due to higher energy → higher fraction of one hit clusters with beam as compared to Sr90

Charge sharing → Due to smallest angular spread → higher fraction of one hit clusters with beam

Effects on reconstructed charge:

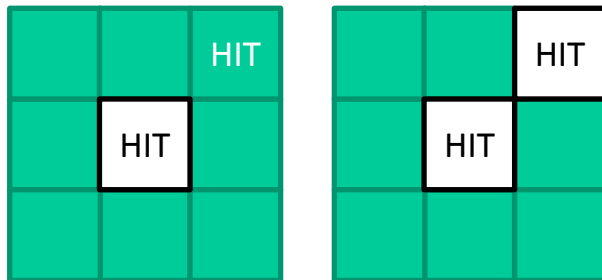
Hit detection only if $V_q > V_{thr}$

→ The bigger clusters have a higher fraction of not detected charge which is below threshold (more charge is not detected)

→ The reconstructed charge is expected to be less than the prediction for both sources

→ The effect is expected to be more pronounced for the Sr 90 measurement

1. Start with a cluster of size one.
2. Check if neighbours to cluster of one pixel distance are hit
 - Yes: add them to cluster → 2.
 - No: nothing → 3.



3. Store cluster and remove all clustered hits from hitmap
4. Start with 1.) until no hits are left

```
def clustering(data=None, pdf_name=None):
    """ Cluster data and return clustered data with one pixel per cluster """
    """ data is of shape [frame_number][x_coord][y_coord] """
    """ pdf_name is the full path to the pdf output file """

    """ Make copy of data for plotting """
    data_raw = np.copy(data)

    print 'Shape of data to be clustered = ', data.shape
    """ clusters is a one dimensional array and holds the number of clusters versus clustersize """
    clusters = np.zeros(shape=(64))
    """ cluster_pixel_data_for_thr_calib is output array containing clustered frames """
    cluster_pixel_data_for_thr_calib = np.zeros(shape=(data.shape[0], data.shape[1], data.shape[2]))

    with PdfPages(pdf_name + '.pdf') as pdf_name:

        """ Loop over all frames """
        for it in range(0,data.shape[0]):
            print 'Frame it ..... ', it, '/', data.shape[0]
            print 'Frame[it] = \n', data[it]
            coords = np.where(data[it][:][:] == 1)
            x = coords[0]
            y = coords[1]
            cluster_count = 0
            print 'x = ', x
            print 'y = ', y

            """ Start to search for all clusters within one frame """
            """ Go over all hit x coordinates, make cluster and remove these from the hit x coordinates and then start over until nothing left """
            while len(x)>0:

                cluster_count += 1
                """ Define the first entry of the array holding all x coordinates of hit pixels to be the start for clustersearch """
                x_y_cluster = [(x[0],y[0])]
                print 'Start values of cluster ', cluster_count, ' = ', x_y_cluster

                keep_searching = True

                iti = 0
                while keep_searching is True:
                    print 'Number of hits to be clustered = ', len(x)
                    iti += 1
                    start_len = len(x_y_cluster)
                    print 'Iterator of while loop = \t', iti, '\t Clustersize of current cluster candidate = \t', start_len
                    for c_it in range(0,len(x_y_cluster)):
                        print 'c_it = ', c_it
                        for x_it in range(0,len(x)):
                            x_diff_cluster_member = x_y_cluster[c_it][0]-x[x_it]
                            y_diff_cluster_member = x_y_cluster[c_it][1]-y[x_it]
                            if (x_diff_cluster_member == 0 or x_diff_cluster_member == 1 or x_diff_cluster_member == -1) and (y_diff_cluster_member == 0 or
                            y_diff_cluster_member == 1 or y_diff_cluster_member == -1):
                                # Here all neighbours are found, also those which already belong to cluster
                                x_y_cluster.append((x[x_it],y[x_it]))

                            x_y_cluster = sorted(set(x_y_cluster)) # remove duplicates from cluster data
                            print 'Iterator of while loop = \t', iti, '\t Clustersize after search = \t', len(x_y_cluster)
                            if start_len == len(x_y_cluster):
                                keep_searching=False
                            if len(x_y_cluster) == len(x):# means all hits of this frame belong to only one cluster
                                keep_searching=False
                            print 'Keep searching bool = ', keep_searching

                print 'Finished a cluster with coordinates = ', x_y_cluster

                """ Cluster finished, store it """
                clustersize = len(x_y_cluster)
                clusters[clustersize] = clusters[clustersize] + 1
                cluster_pixel_data_for_thr_calib[it][x[0]][y[0]]+=1

                print 'Clustered frame[it] = \n', cluster_pixel_data_for_thr_calib[it]

                """ Remove clustered hits from data """
                for hit in range(0,len(x_y_cluster)):
                    data[it][x_y_cluster[hit][0]][x_y_cluster[hit][1]]=0
                print 'Mod hitmap after reduction of hits from ', cluster_count, ' clusters = ', data[it]

                """ Make new x and y hit coordinates for next cluster """
                print '... continue with next cluster'
                coords = np.where(data[it][:][:] == 1)
                x = coords[0]
                y = coords[1]
                print 'next x', x
                print 'next y', y

            if clustersize > 3:
                plotting.plot_2D_hitmap(hist2d=data_raw[it], title='Occupancy map (V2) it' + str(it) + ' clusters = ' + str(cluster_count)+ ' clustersize = '
                + str(clustersize), zlabel='Hits', z_max=None, filename=pdf_name, savename='test')

            print 'Found clusters: ', cluster_count
            if cluster_count > 1:
                plotting.plot_2D_hitmap(hist2d=data_raw[it], title='Occupancy map (V2) it' + str(it) + ' clusters = ' + str(cluster_count), zlabel='Hits',
                z_max=None, filename=pdf_name, savename='test')

            plotting.plot_2D_hitmap(hist2d=data_raw[it], title='Cluster distribution 6x6 V2', filename=pdf_name, savename='test')
            print 'Clustering done.'
```

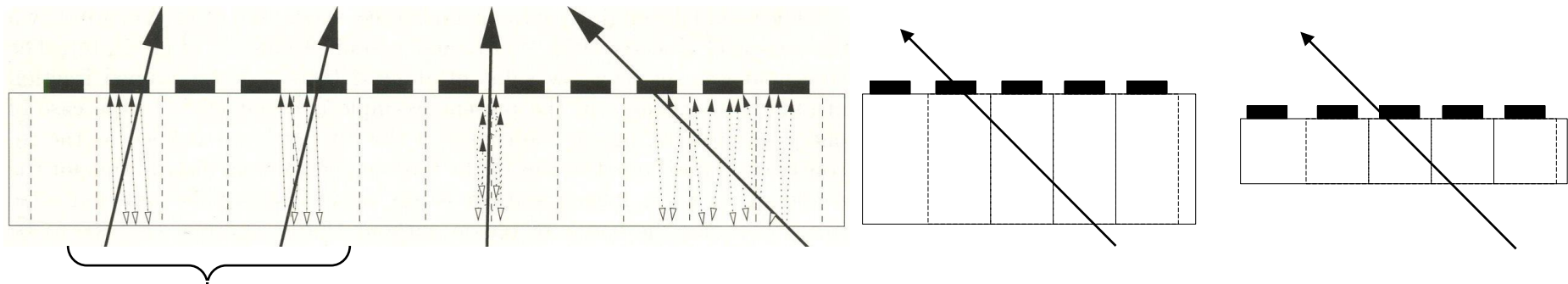
Charge sharing (geometrical effects)

Charge cloud spreads laterally due to diffusion and therefore charge can be shared between neighbouring pixels → clusters

Charge sharing is geometrically affected by:

- Position of track with respect to sensor surface
- Geometry of pixel cell
- Angle of track with respect to sensor surface

Electrons: ↑
 Holes: ↓
 Incident particle: ↑



Same angle but different positions

→ Left: Clustersize: 1

→ Right: Clustersize: 2

Track exactly on pixel edge

Clustersize: 2

Clustersize: 3

Clustersize: 3

Clustersize: 2

- Position of track with respect to sensor surface → Not known, same for Sr90 and electron beam
- Geometry of pixel cell → almost cubic, $40\ \mu\text{m} \times 40\ \mu\text{m} \times 50\ \mu\text{m}$, fixed
- **Angle** of track with respect to sensor surface → Different for Sr 90 and electron beam

→ The bigger the angular spread, the higher the fraction of bigger clusters.

→ The smaller the angular spread, the higher the fraction of one hit clusters.

Angular spread of Sr 90 > Angular spread Sr 90 collimated > focussed electron beam

Expectation:

→ Highest fraction of one hit clusters for electron beam measurement

→ Smallest fraction of one hit clusters with uncollimated Sr 90 source

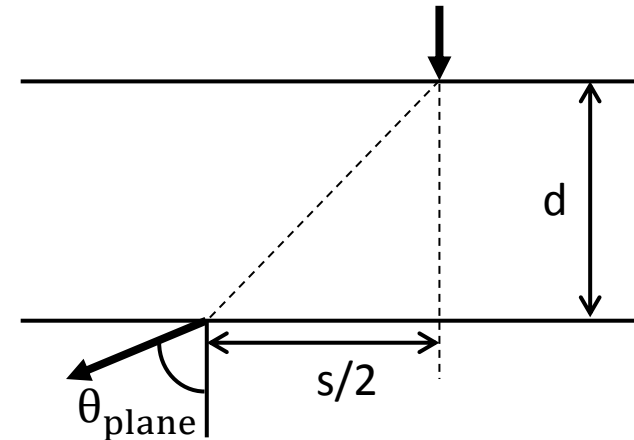
Multiple scattering

Deflection of particle in matter by coulomb scattering, can be calculated:

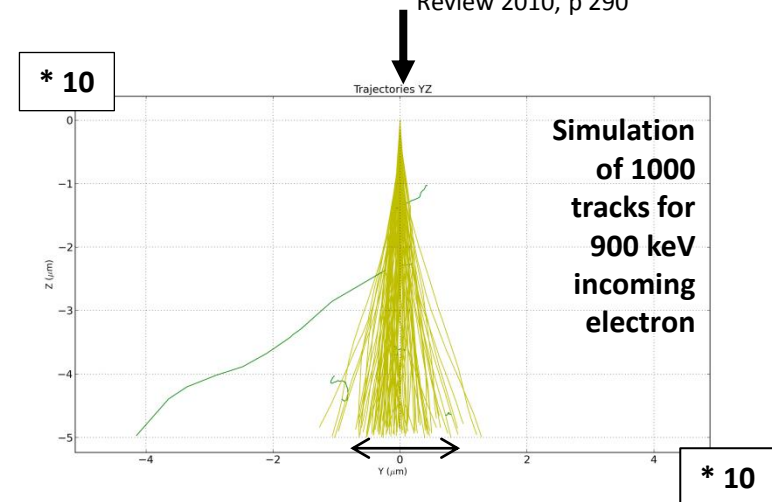
- $\theta_{\text{plane}} = \frac{13.6}{\beta p [\text{MeV}/c]} z \sqrt{\frac{x}{X_0}} \left(1 + 0.038 \ln \frac{x}{X_0} \right)$
- θ_{plane} = sigma of Gauss fit to central 98% of the scattering angles [3]
- $s/2 = d \cdot \tan \theta_{\text{plane}} \approx d \cdot \theta_{\text{plane}}$; $d = 50 \mu\text{m}$
- $s(1 \text{ sigma}) = 20 \mu\text{m}$ for Sr 90 (2.28 MeV)
- $s = 0.003 \mu\text{m}$ for 3.2 GeV electron beam

Simulation of electron trajectories for 2.28 MeV incident particle agrees with calculation

→ Expectation: Higher fraction of one hit clusters with electron beam measurement as compared to Sr 90 measurement



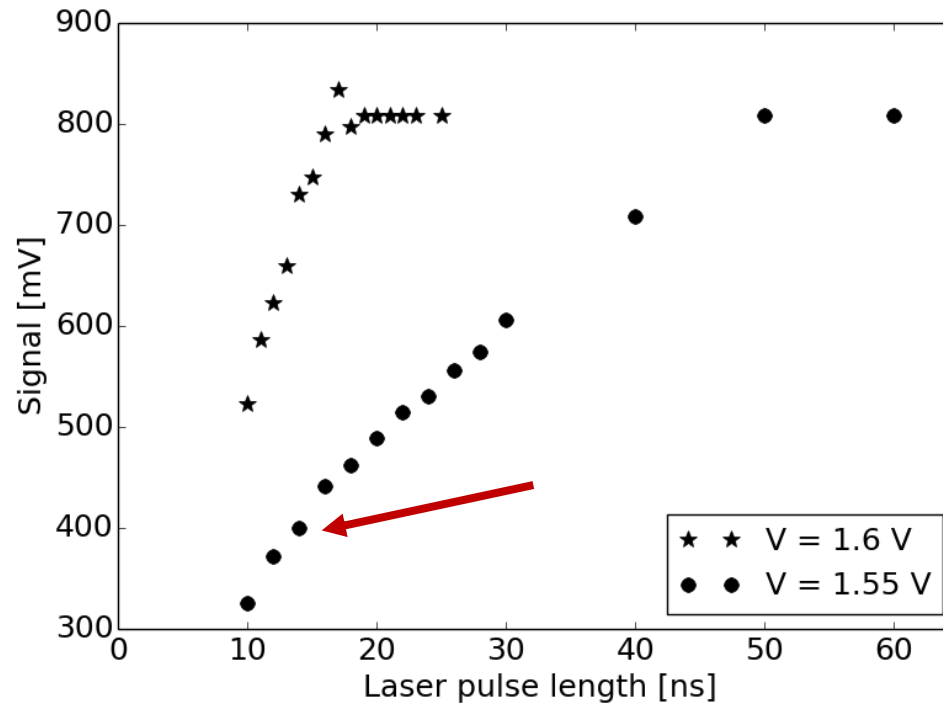
[3] Bichsel, Groom, Klein, PDG Review 2010, p 290



- β, p, z properties incoming particle
- $\frac{x}{X_0}$ traveled distance in units of radiation length

LASER measurements

- Injection of laser can be configured by voltage and laser pulse width
- Linearity measured with complete system (Laser + EPCB01)

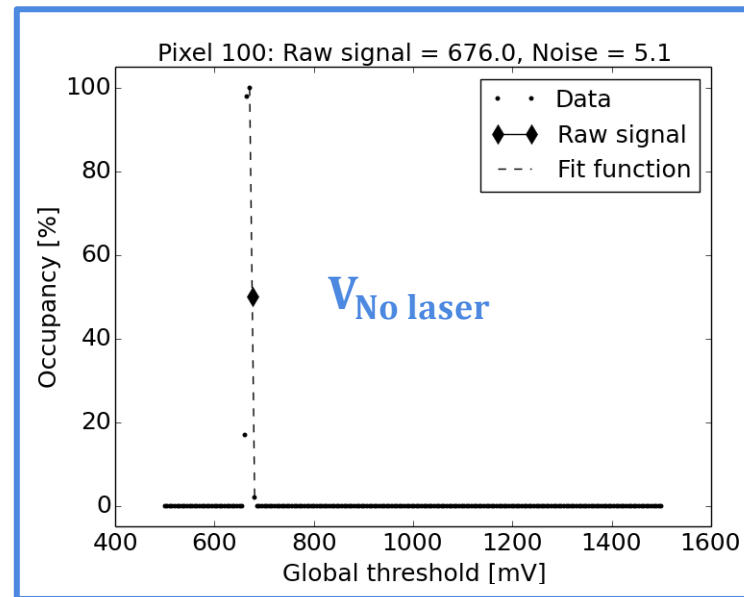
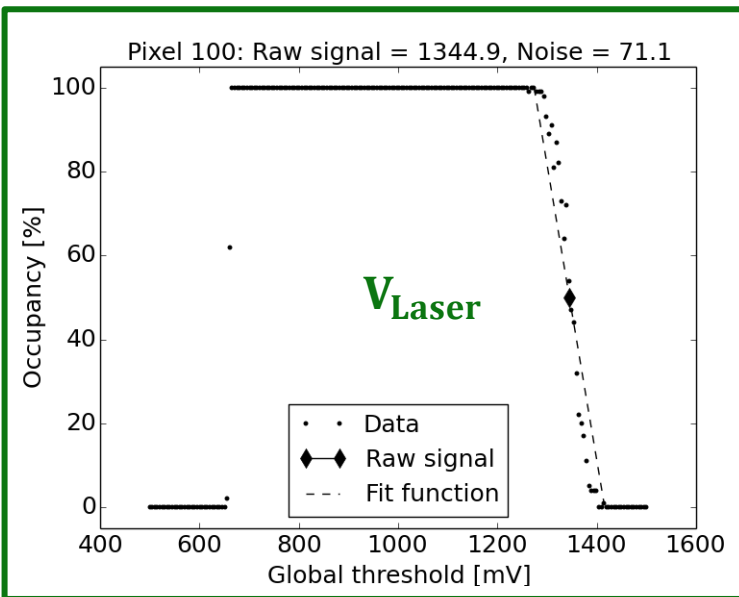
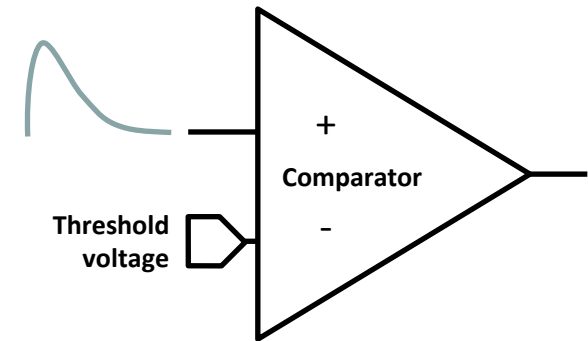


→ Typical setting: 14 ns, 1.55 V (linear region)

- Move laser over matrix and measure the signal

Signal:

- Threshold scan with laser as injection
- Scan range [500,1500] mV to cover signal and baseline
- V_{Laser} and o_{Laser} determined from line fit

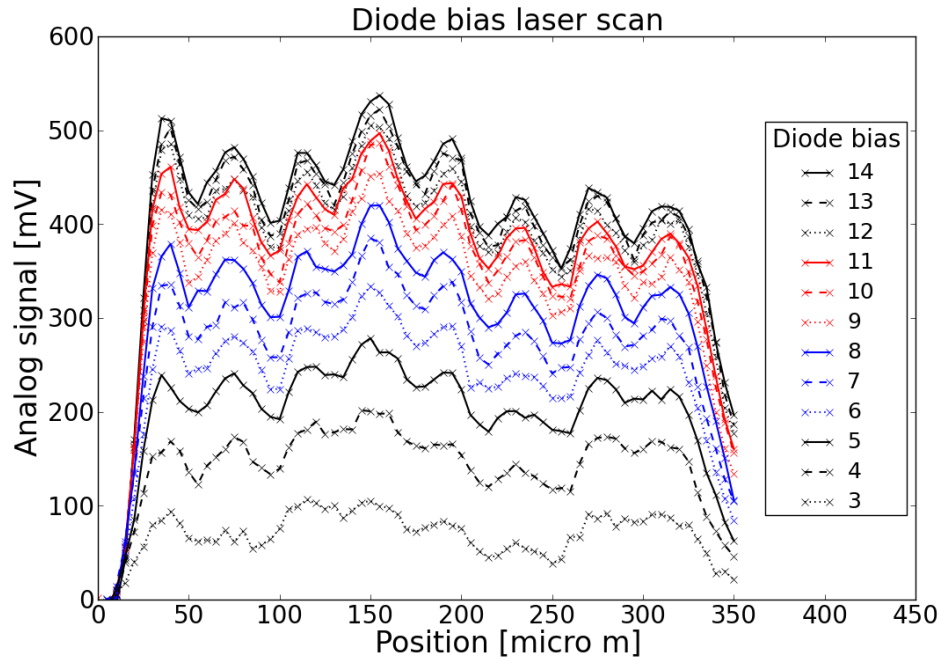


Noise cut:
 $V_{Signal} > V_{Cut}$
 $V_{Cut} = 3 \cdot o_{No\ laser}$

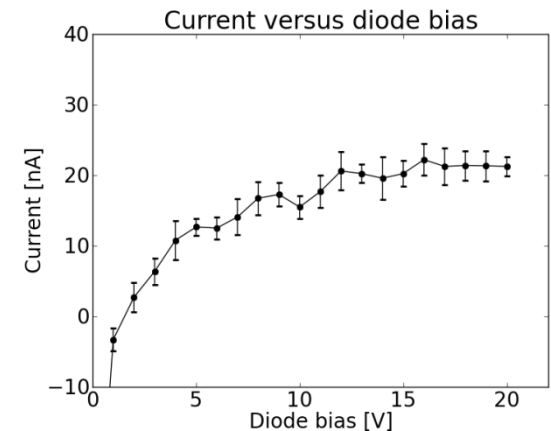
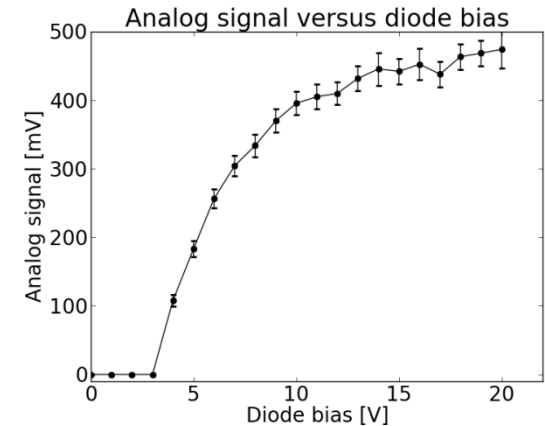
$$V_{Signal} [mV] = (V_{Laser} [mV] - V_{No\ laser} [mV]) \longrightarrow V_{Signal} [mV] = 669\ mV$$

- Scan the same column with the laser for different diode bias settings

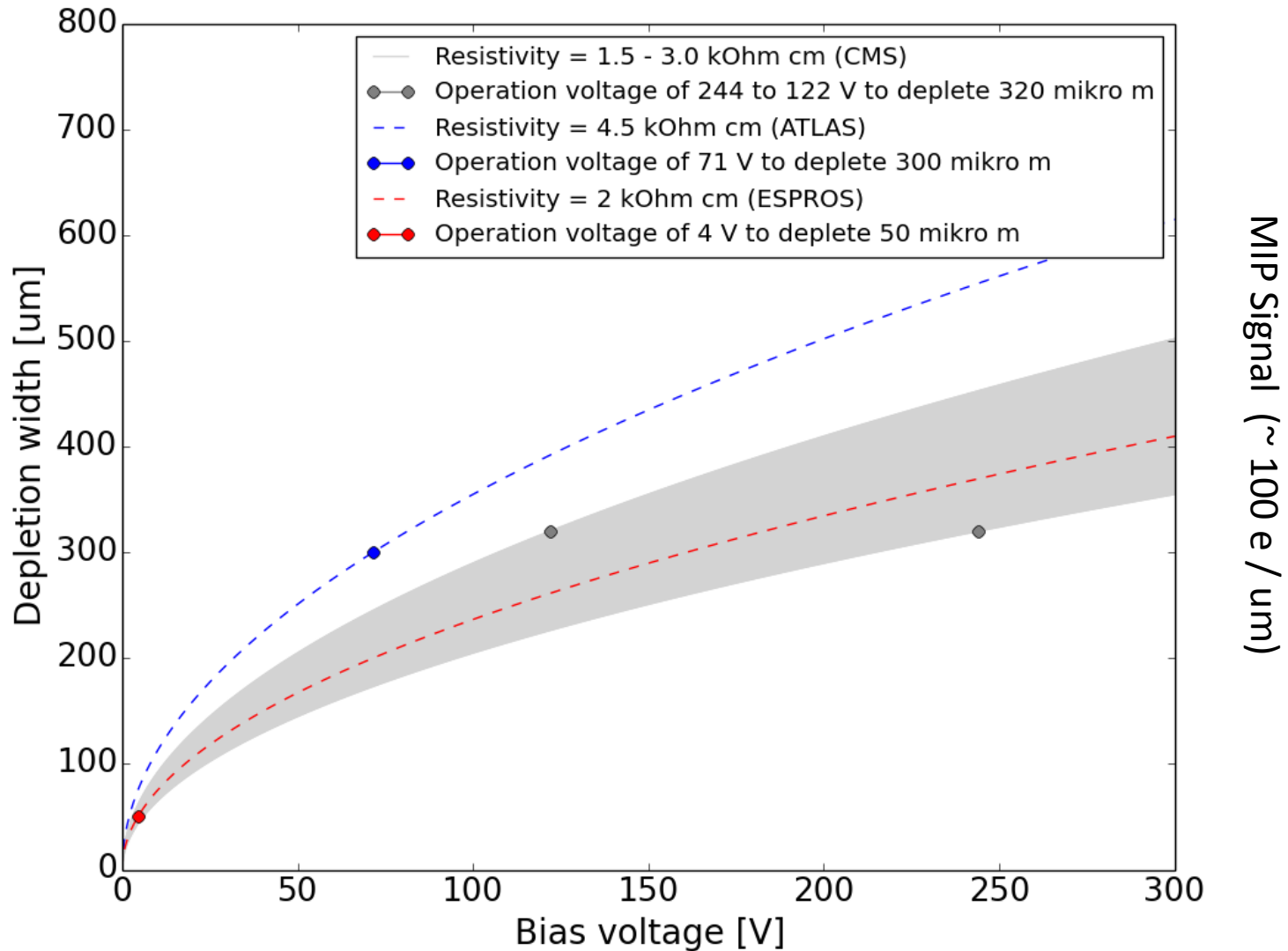
For one pixel and fixed laser position:



- Increasing diode bias voltage results in higher signals, as expected
- Saturation of signal above 10 V (→ depleted sensor)
- Leakage current ~ 20 nA

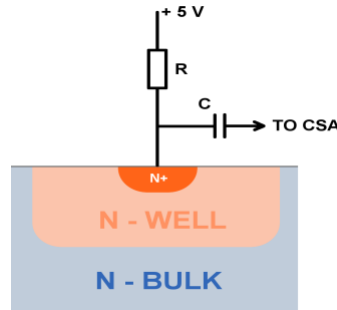


Depletion widths for different resistivities

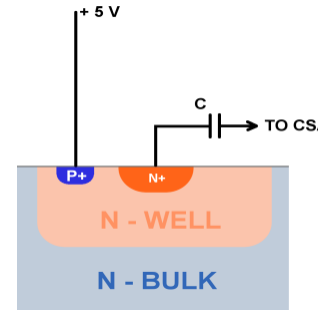


- Bias options:

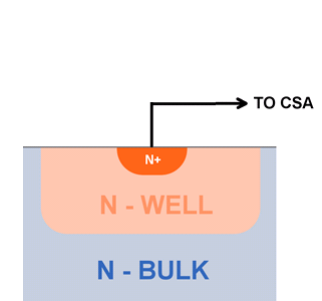
AC coupling
Resistor bias



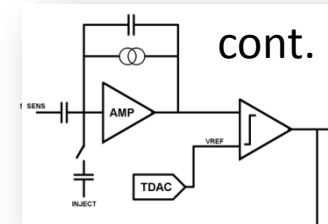
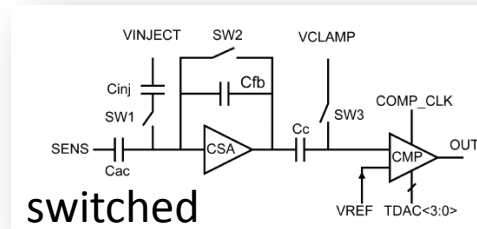
AC coupling
Diode bias



Direct coupling
Direct bias



- Pixel electronics options:

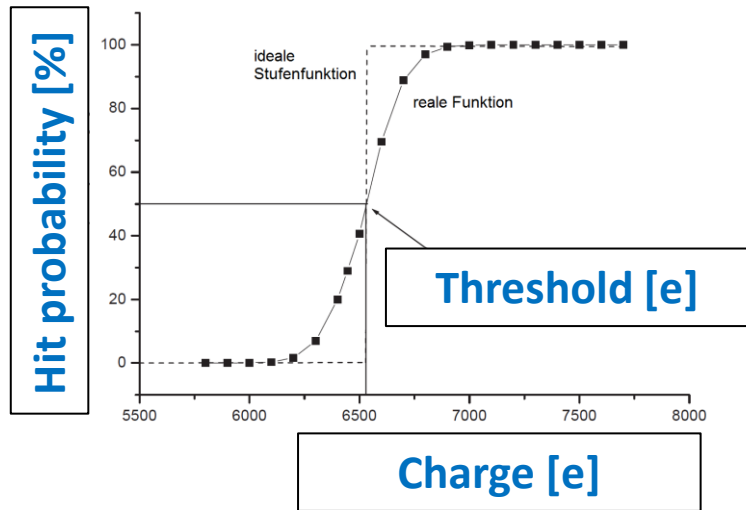


- Matrix versions:

Lab test results:

- FE characterization of all matrices
- Sensor characterization of diode bias only

Matrix version	Biasing and coupling	Analog FE
V1	Resistor + AC	Continuous
V2	Diode + AC	Continuous
V3	Direct + DC	Continuous
V4	Direct + DC	Switched
V5	Diode + AC	Switched
V6	Resistor + AC	Switched



Threshold

- Injection of different charges
- Plot hits versus charge injection
- Ideal case: step function
- Noise \rightarrow Scurve (step + gauss)
- Threshold: x – value of fit function at y = 50 %
- $P_{\text{Treffer}}(Q) = \Theta(Q - Q_{\text{Schwelle}}) \cdot \exp\left(-\frac{Q^2}{2\sigma^2_{\text{Rauschen}}}\right)$

Amplitude

- Injection of the same charge
- Variation of the threshold
- Plot hits versus threshold
- Ideal case: step function
- Noise \rightarrow Scurve (step + gauss)
- Amplitude: x – value of fit function at y = 50 %
- $P_{\text{Treffer}}(Q) = \Theta(Q - Q_{\text{Schwelle}}) \cdot \exp\left(-\frac{Q^2}{2\sigma^2_{\text{Rauschen}}}\right)$

