

# Accelerating the development and utilization of new-type *Brassica napus* gene pool by genomics-based approaches

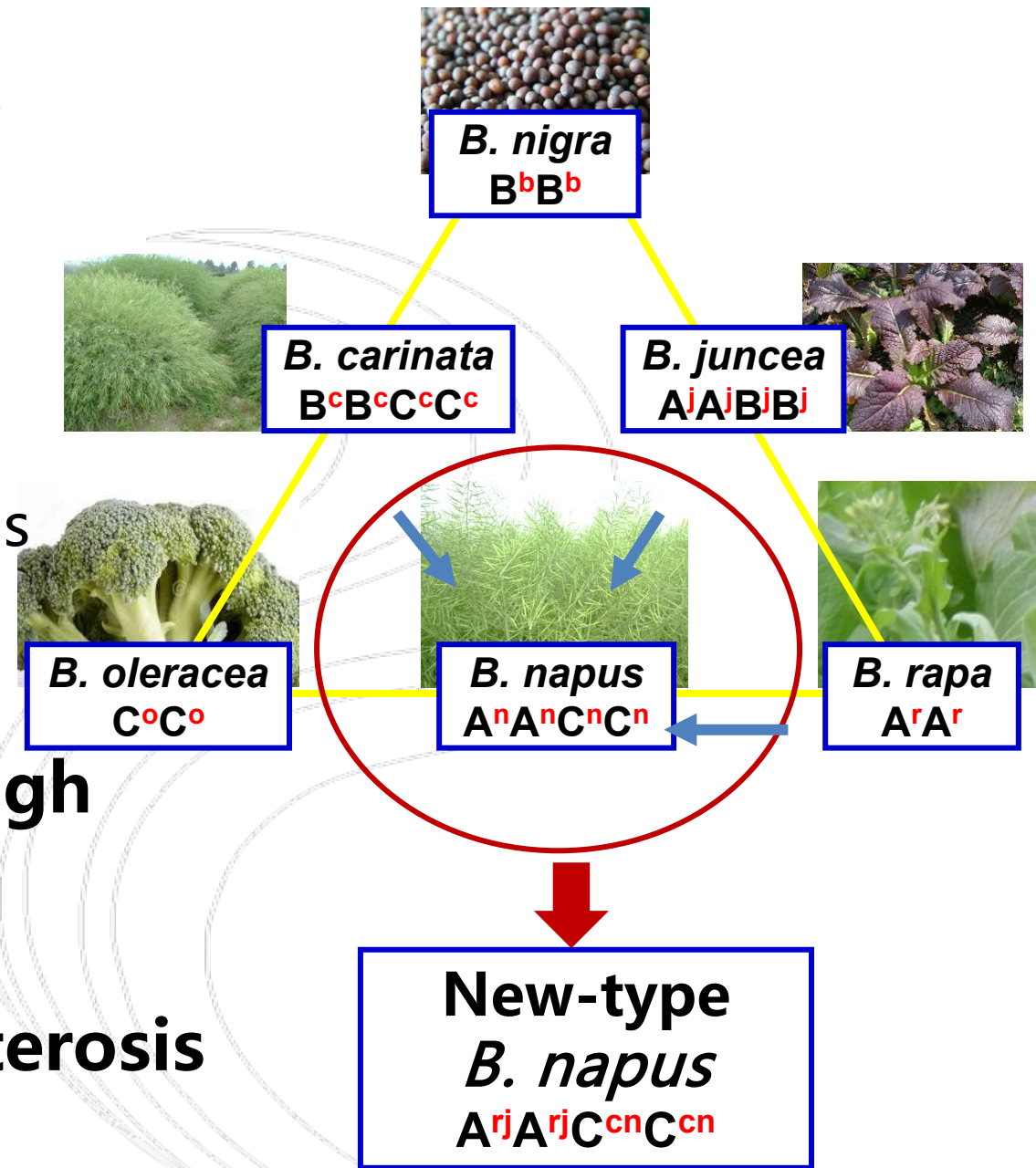
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**Sydney**

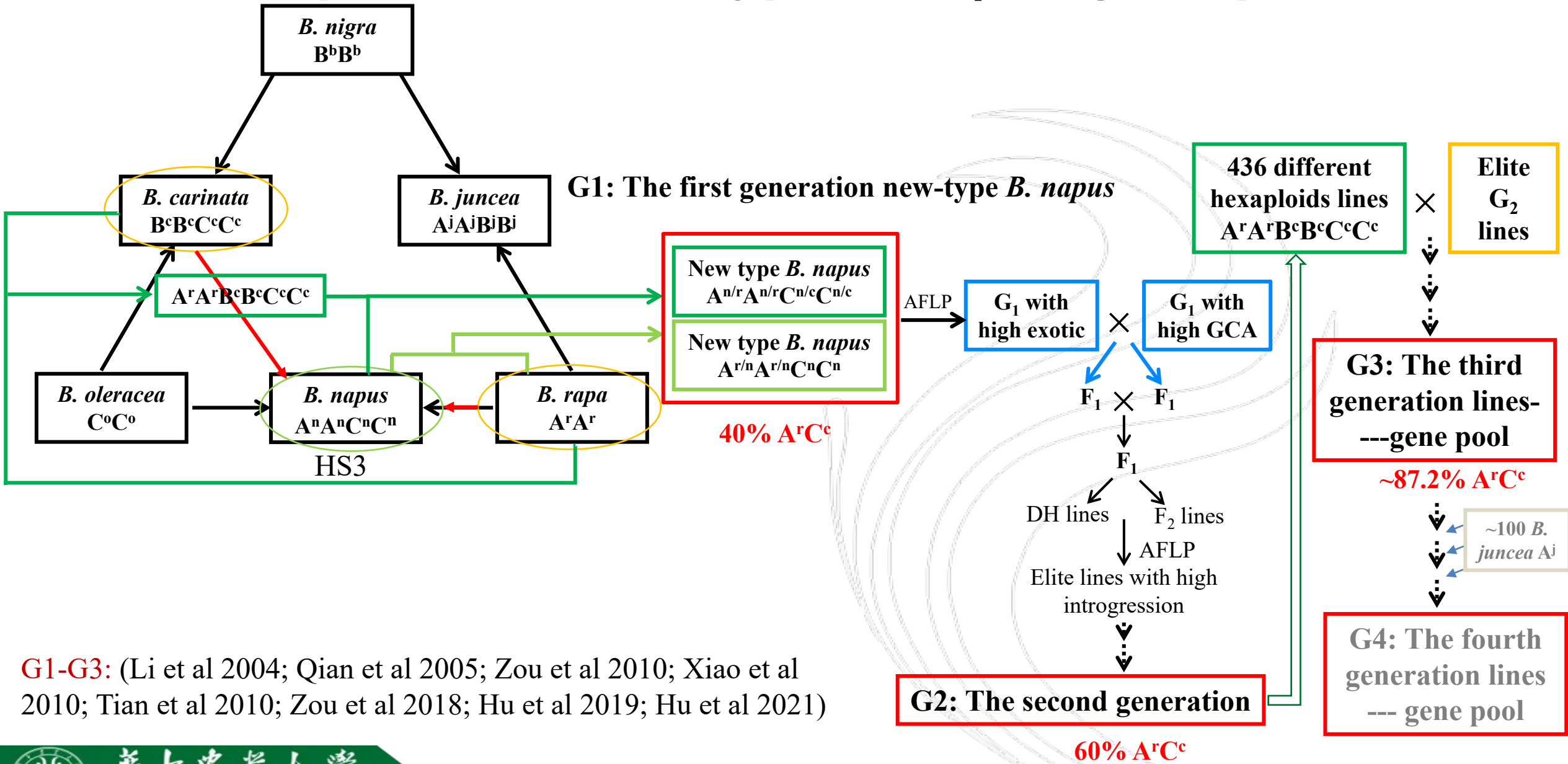
**26. 09. 2023**

- *Brassica napus* is an important oilseed crop and heterosis utilization is most important approach for rapeseed production
- However, it has narrow genetic diversity which also affects the utilization of heterosis



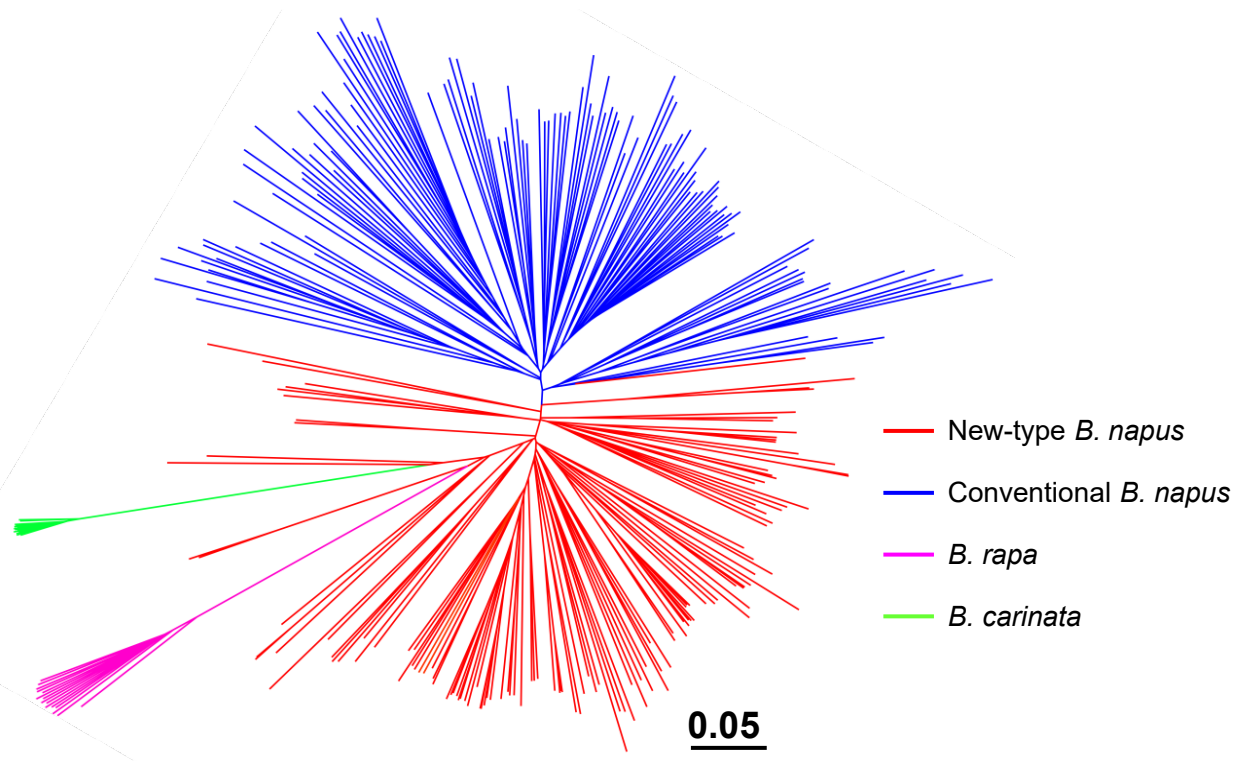
- ◆ Broadening its genetic basis through subgenome reconstruction
- ◆ Utilization of intersubgenomic heterosis

# The development of new-type *B. napus* gene pool

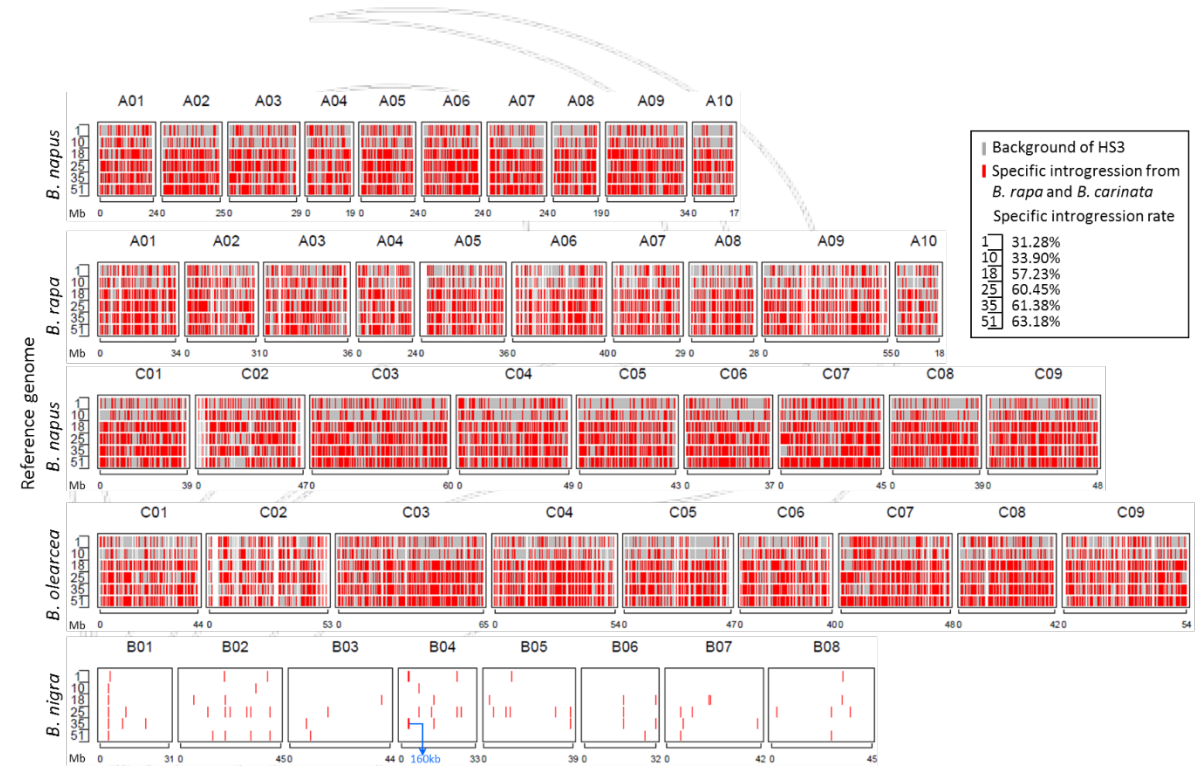


G1-G3: (Li et al 2004; Qian et al 2005; Zou et al 2010; Xiao et al 2010; Tian et al 2010; Zou et al 2018; Hu et al 2019; Hu et al 2021)

# New-type *B. napus* is an independent subgroup



Inbred lines evaluated by *Brassica* 60K-SNP array

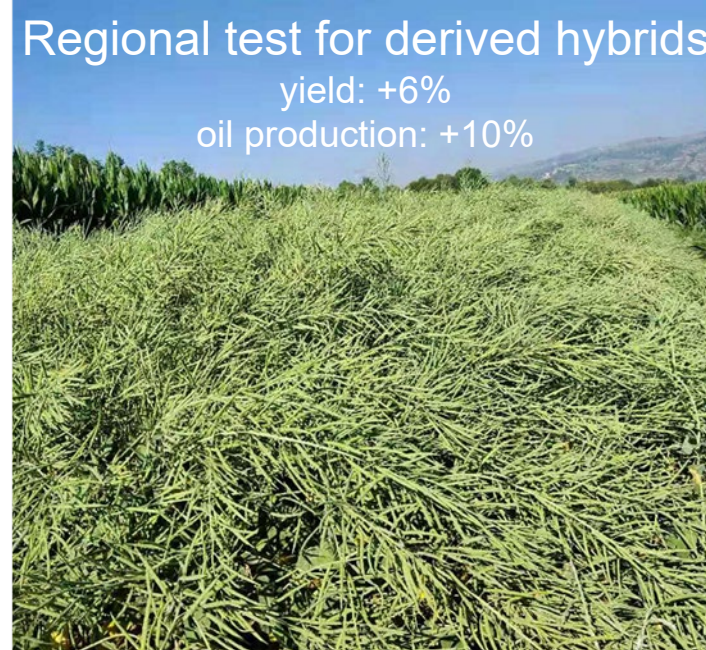


Rich introgression components from *B. napus* relatives

# Excellent variations and breeding application

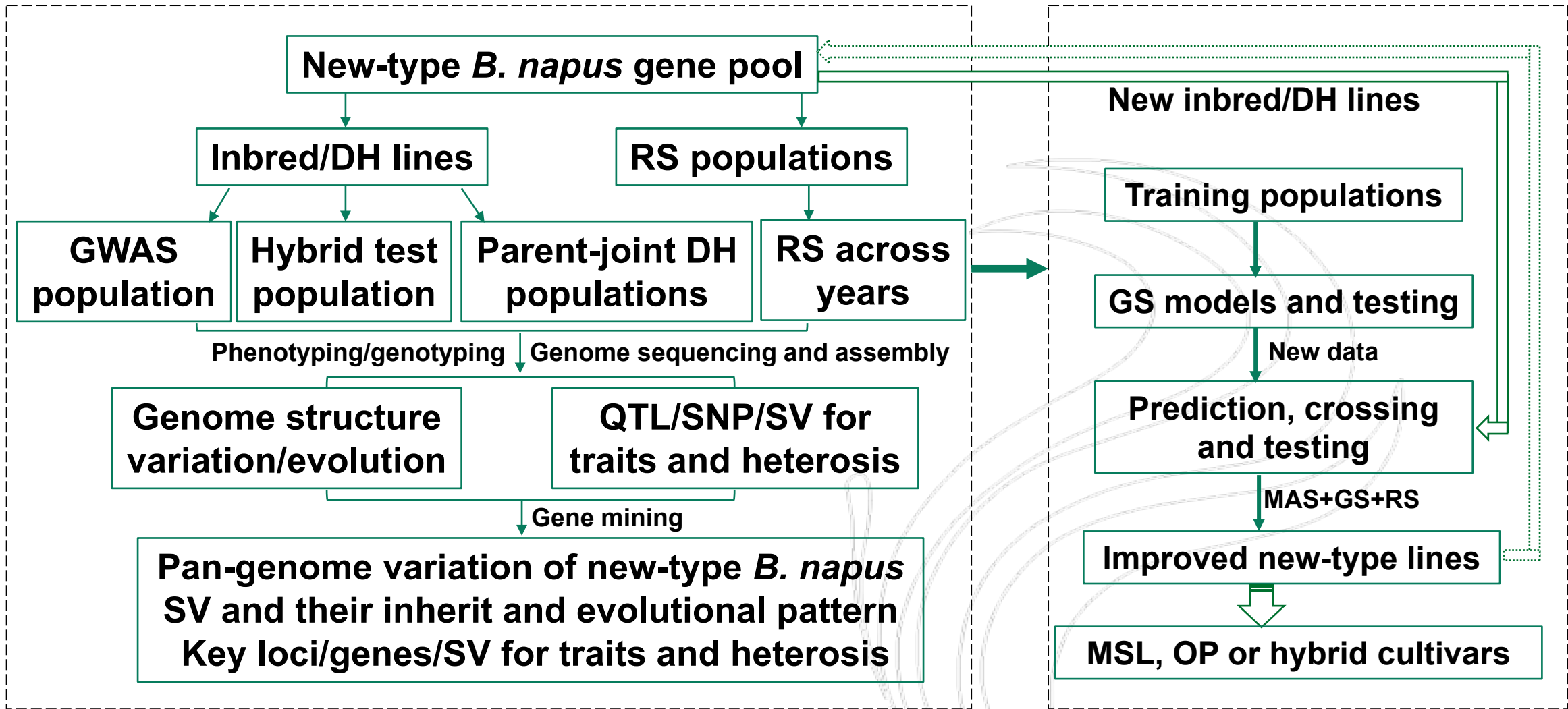


ZS11 (control) 3.50g  
 AM032 7.10g  
 AM140 Dark brown  
 19N940 Brown

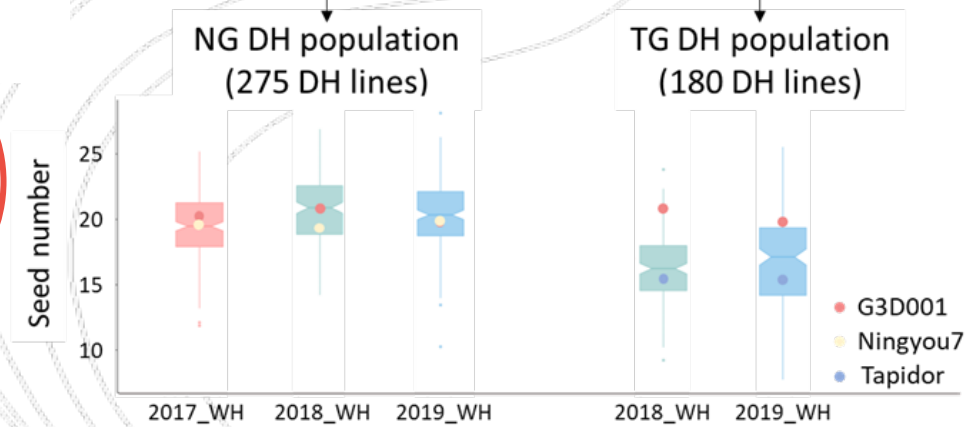
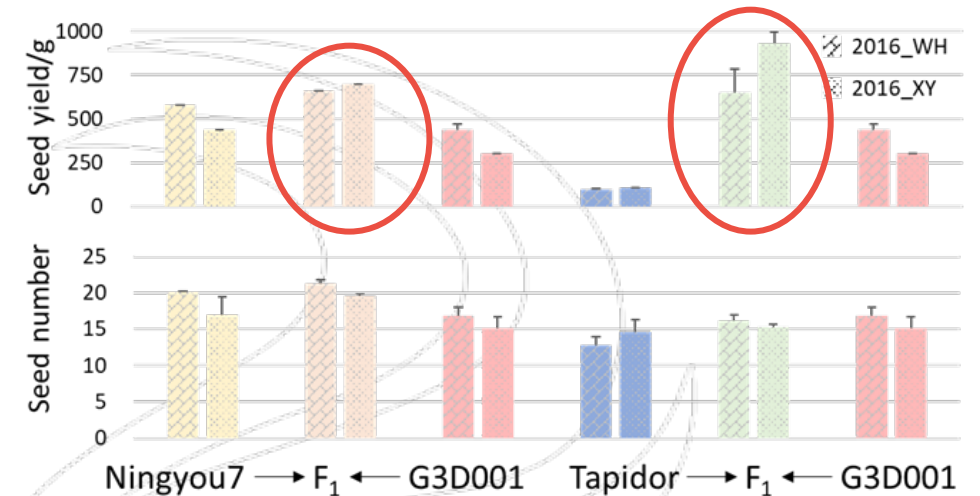
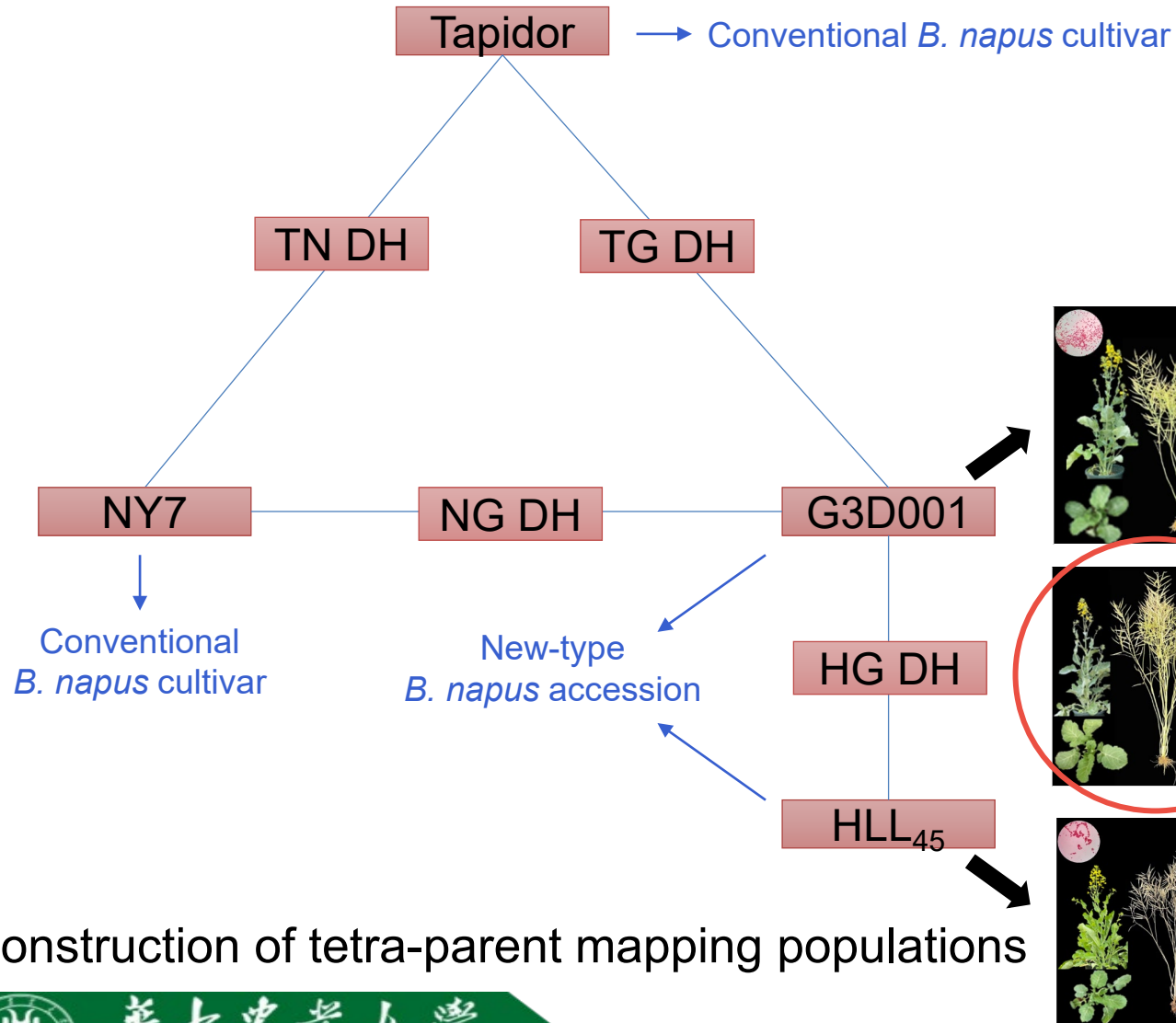


# Questions

- What's the genome-wide novel variation of new-type *B. napus* population and which are favorable for breeding?
- How are these abundant variations fixed following *de novo* domestication and intensive selection?
- How should we accelerate the development of the gene pool and the application of elite variations with genomic tools for practical breeding ?



# Mapping populations for variation identification



Phenotyping of F<sub>1</sub> and DH lines

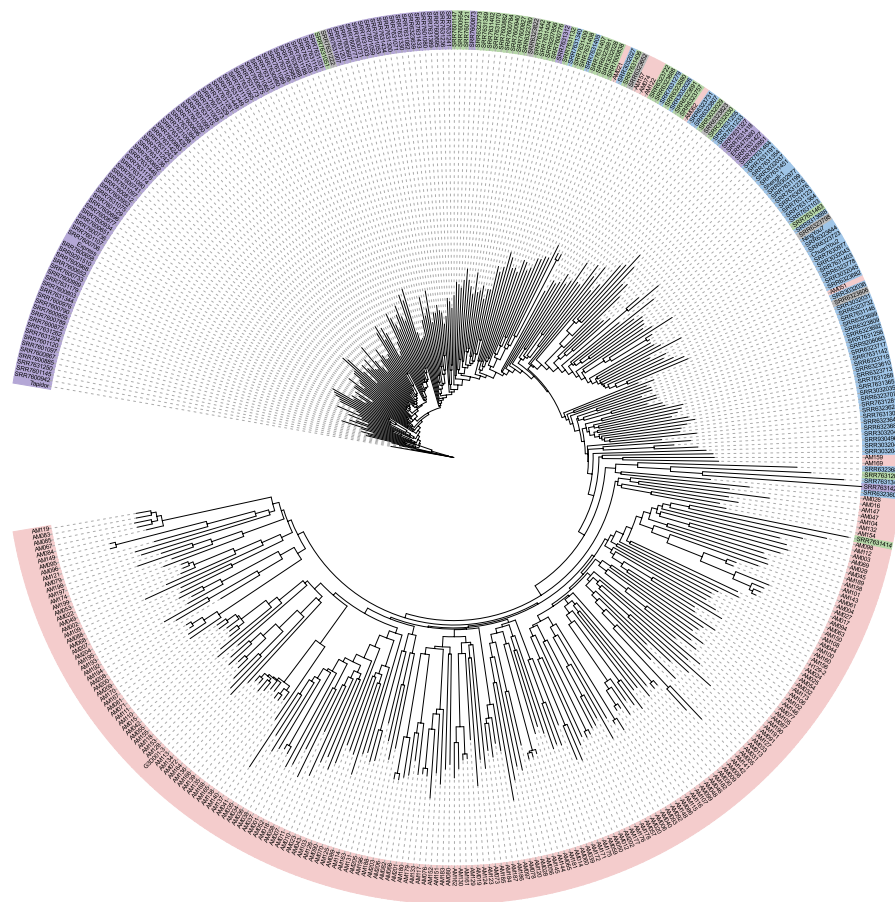
(Hu et al., unpublished data)

# GWAS population: a panel of ~300 diverse lines

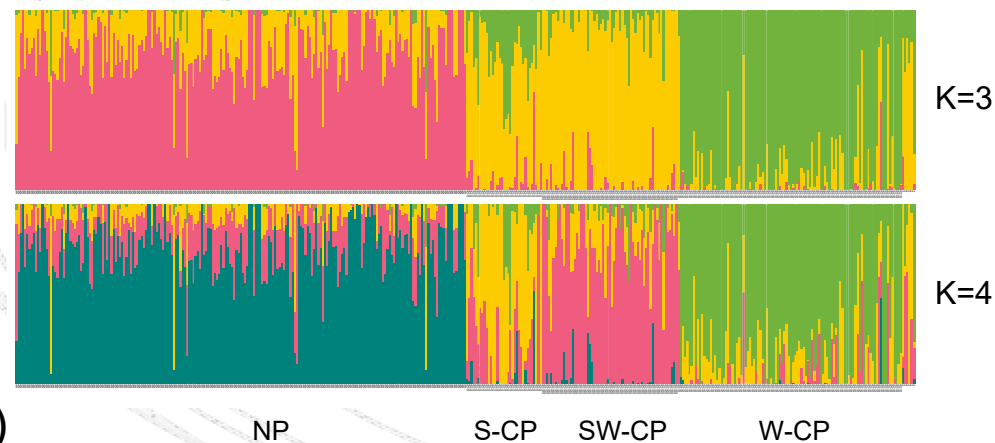
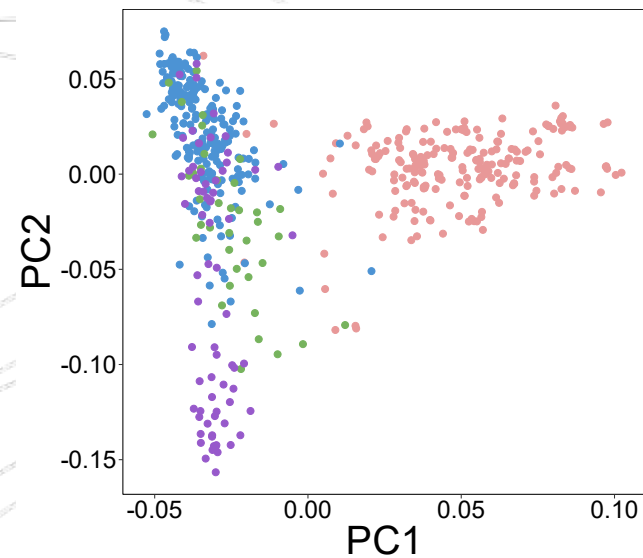


# Progress 1 Comprehensive SV in new-type *B. napus*

## Substantial sub-population differentiation of new-type *B. napus* population (NP) revealed by whole-genome sequencing

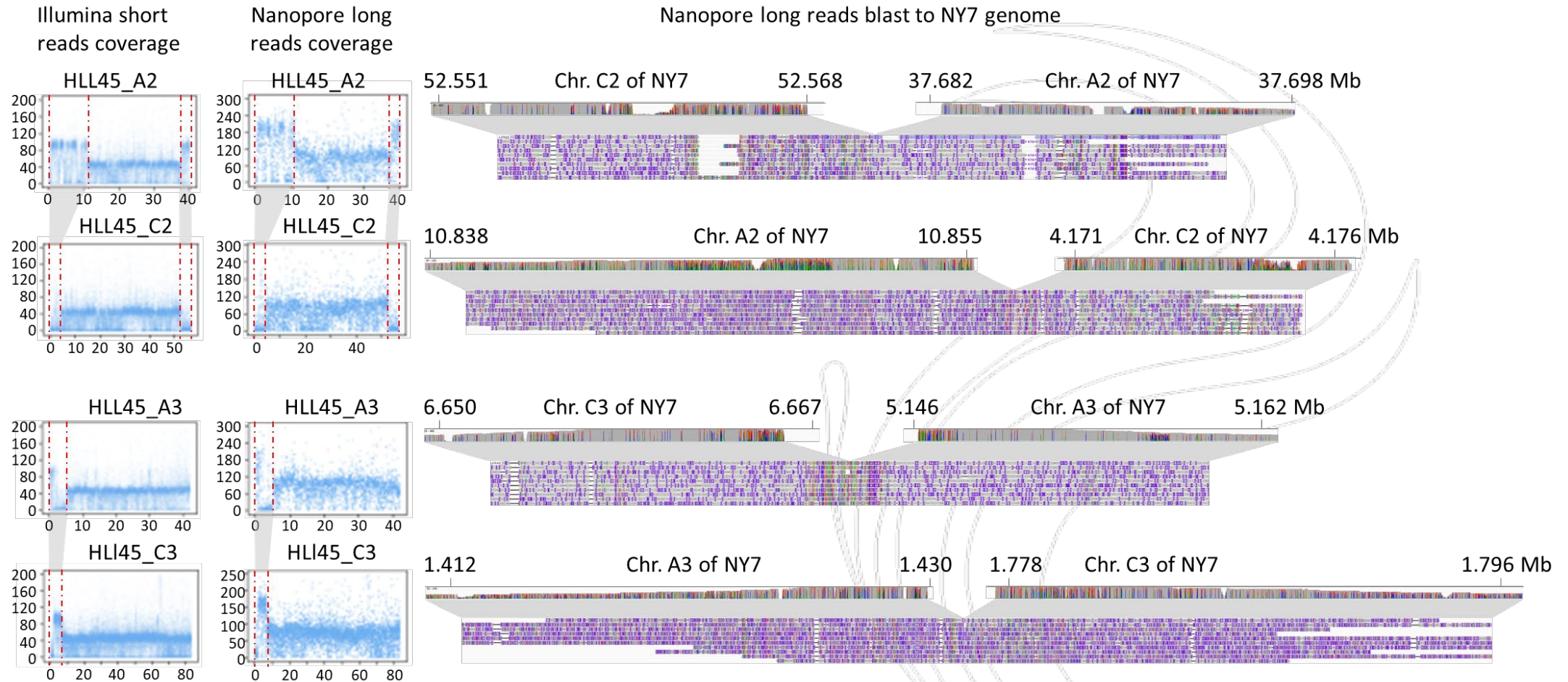


- NP
- S-CP
- SW-CP
- W-CP
- Unknown



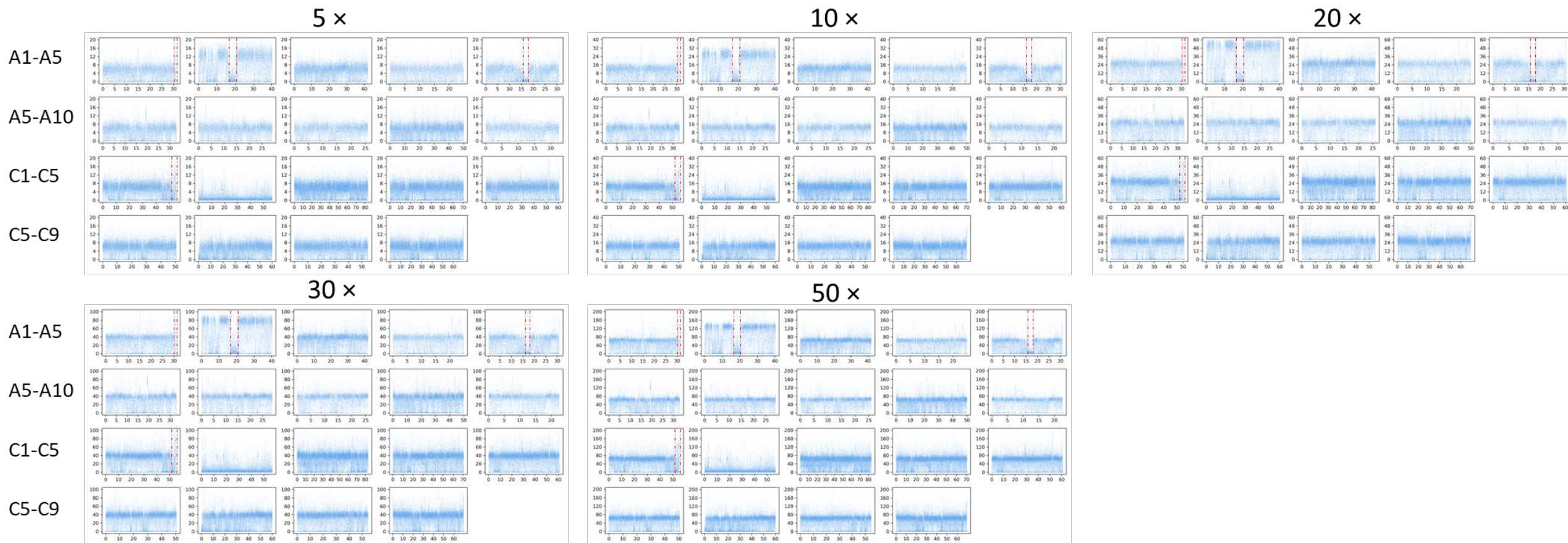
(Han Qin, unpublished data)

# Extensive large-scale (>500 kb) SVs fixed in new-type *B. napus* lines



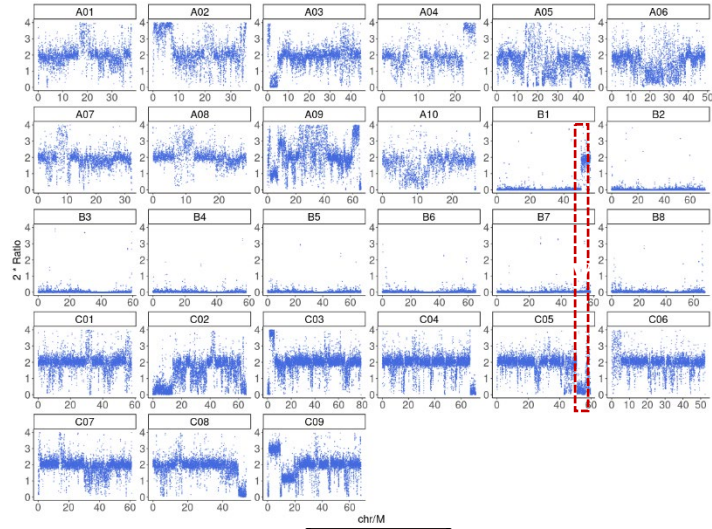
(Hu et al., unpublished data)

The SVs detected by different coverage of sequence data were same, and the SVs detected by  $5\times$  genomic sequence data were reliable.

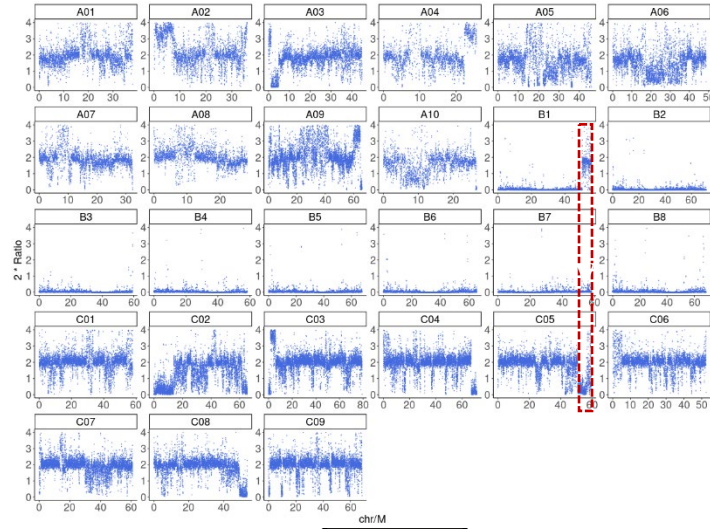


(Unpublished data)

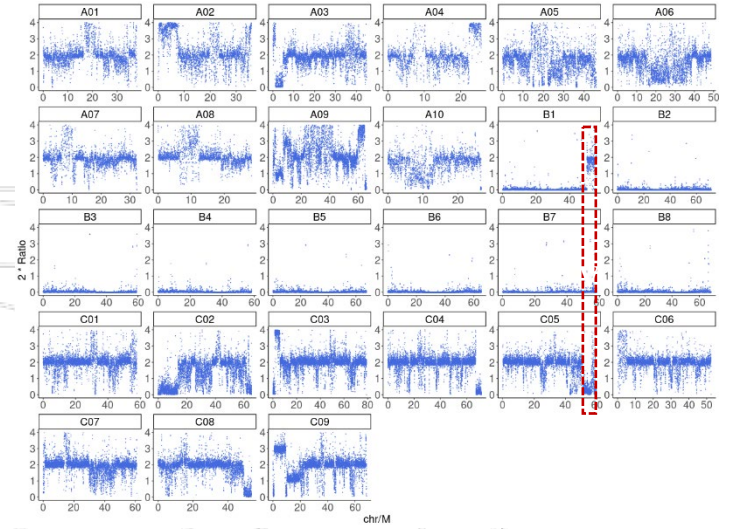
# Stable inheritance of large SVs in new-type *B. napus*: an example on C5



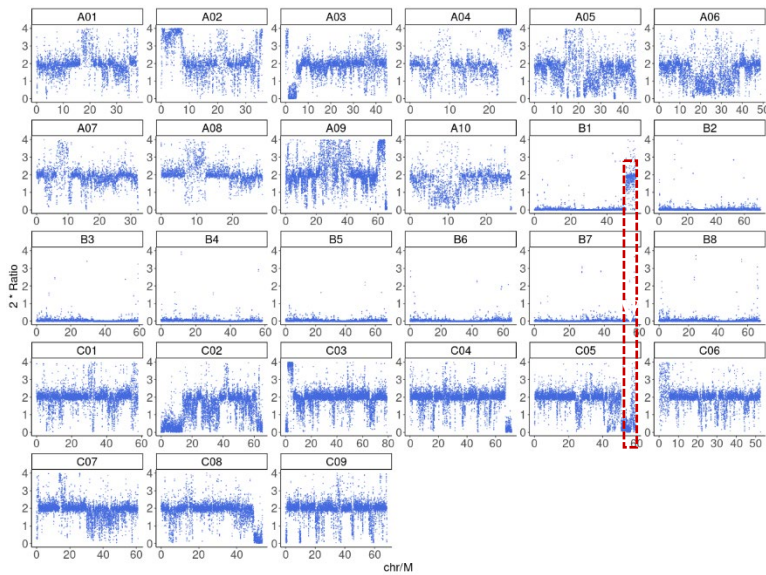
F<sub>12</sub>



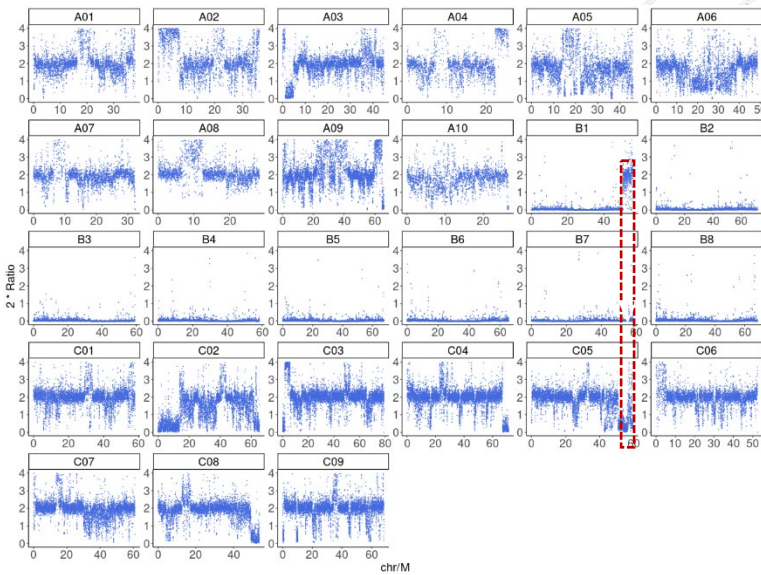
F<sub>14-1</sub>



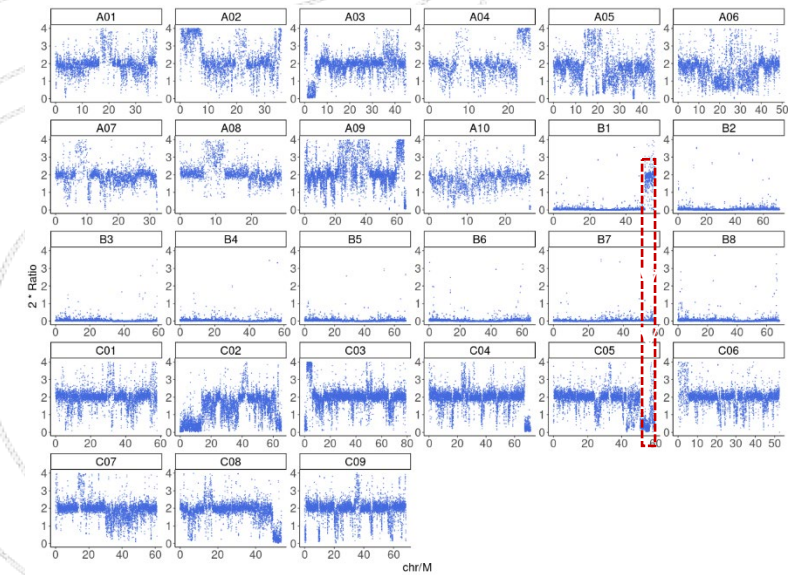
F<sub>14-3</sub>



F<sub>16</sub>



F<sub>17-1</sub>

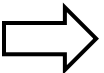


F<sub>17-2</sub>

(Unpublished data)

# Complex pattern of the polymorphism markers presented in the DH mapping populations: SV and exotic introgressions

| Population | Reference genome       | Polymorphic marker |
|------------|------------------------|--------------------|
| HG DH      | $A^n C^n$              | 1,326,409          |
|            | $A^r B^{c/ni} C^{c/o}$ | 468,277            |
|            | Total                  | 1,794,686          |
| NG DH      | $A^n C^n$              | 1,420,556          |
|            | $A^r B^{c/ni} C^{c/o}$ | 326,414            |
|            | Total                  | 1,746,970          |
| TG DH      | $A^n C^n$              | 1,479,851          |
|            | $A^r B^{c/ni} C^{c/o}$ | 564,099            |
|            | Total                  | 2,043,950          |



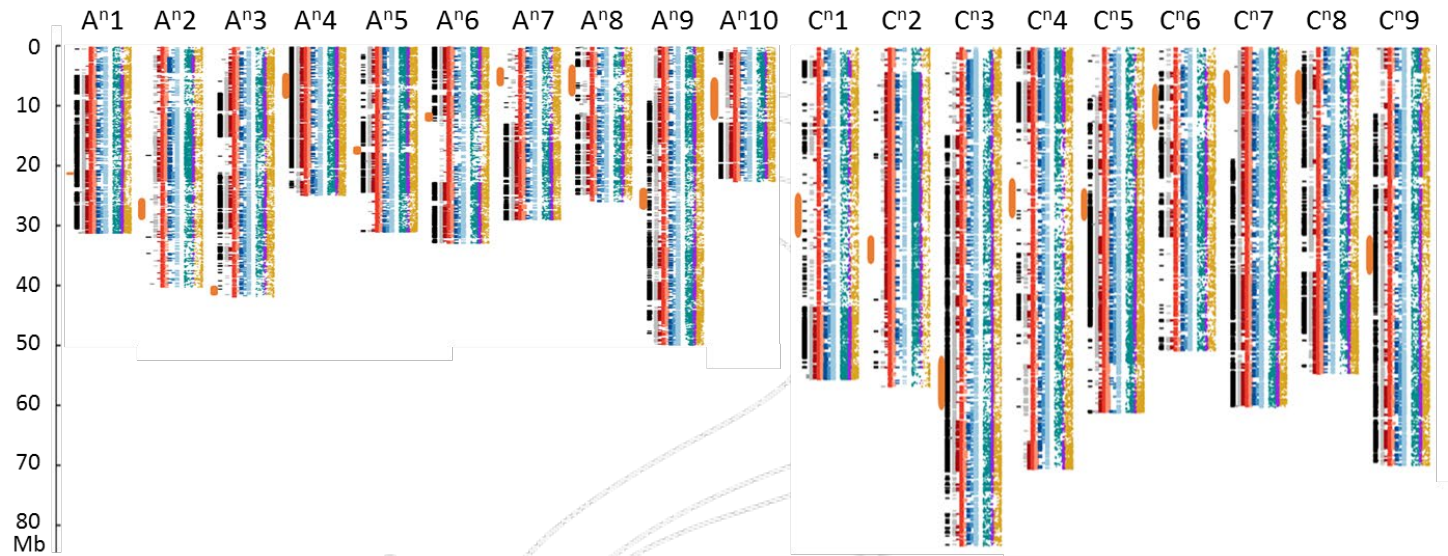
Specific

| Marker type | Parental genotype |        |
|-------------|-------------------|--------|
|             | G3D001            | H129-2 |
| Normal      | 0/0               | 1/1    |
|             | 1/1               | 0/0    |
| PAV         | ./.               | 1/1    |
|             | ./.               | 0/0    |
|             | ./.               | 0/1    |
|             | 0/0               | ./.    |
|             | 0/1               | ./.    |
|             | 1/1               | ./.    |
| Hemi        | 0/0               | 0/1    |
|             | 0/1               | 1/1    |
|             | 0/1               | 0/0    |
|             | 1/1               | 0/1    |

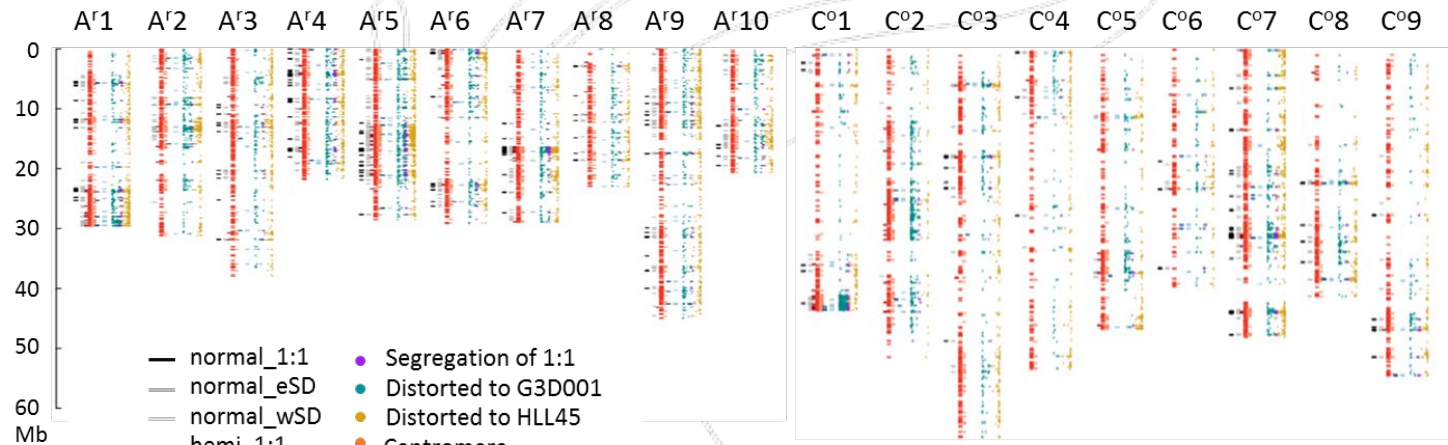
(Hu et al., unpublished data)

# Distribution of various markers across the whole genome

Conventional markers with *B. napus* ( $A^nA^nC^nC^n$ ) genome as the reference



Introgressive markers with *B. rapa* ( $A^rA^r$ ) and *B. oleracea* ( $C^oC^o$ ) genome as the reference



PAV/Hemi markers

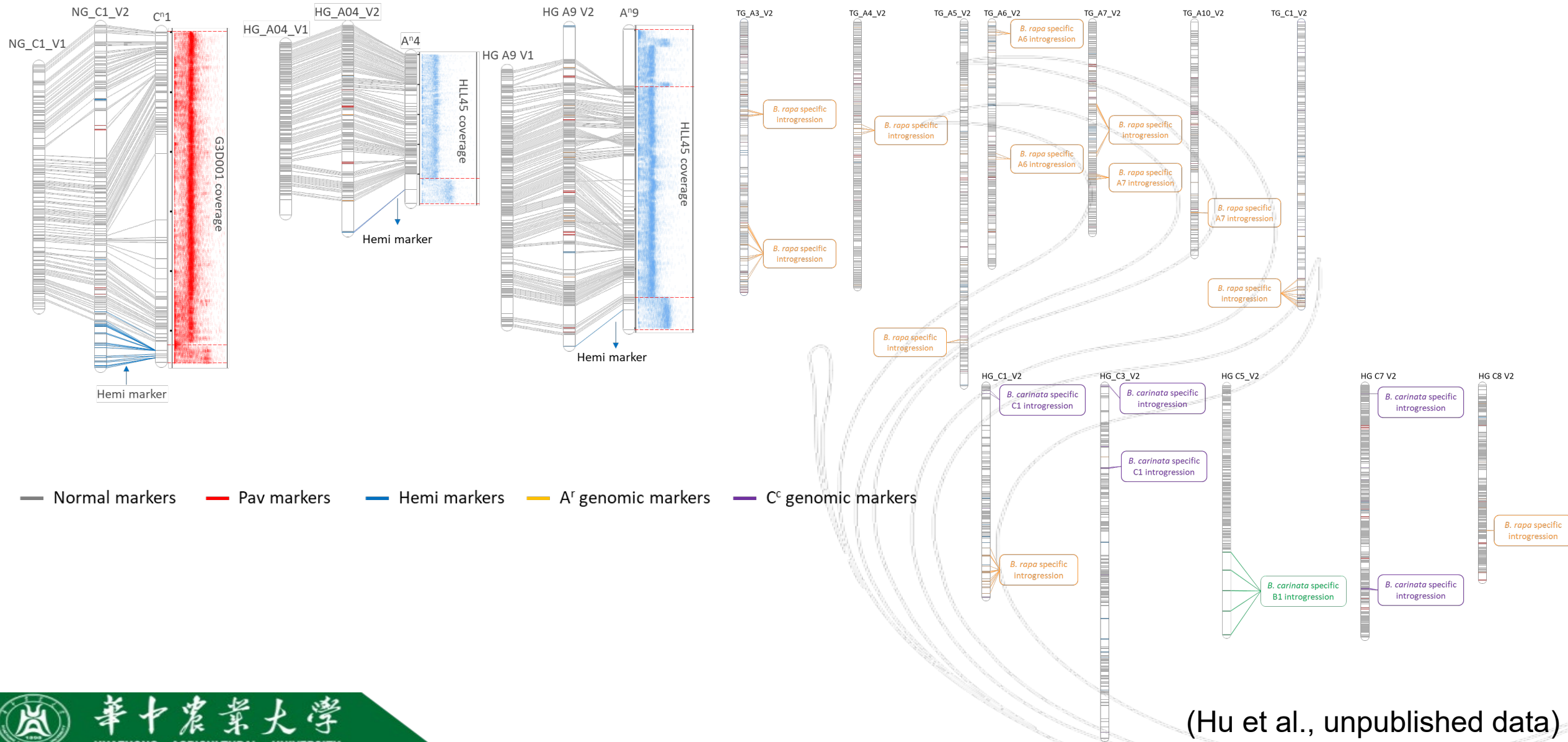
- normal\_1:1
- normal\_eSD
- normal\_wSD
- hemi\_1:1
- hemi\_eSD
- hemi\_wSD
- PAV\_1:1
- PAV\_eSD
- PAV\_wSD
- Segregation of 1:1
- Distorted to G3D001
- Distorted to HLL45
- Centromere

\*eSD: extreme segregation distortion

\*wSD: weak segregation distortion

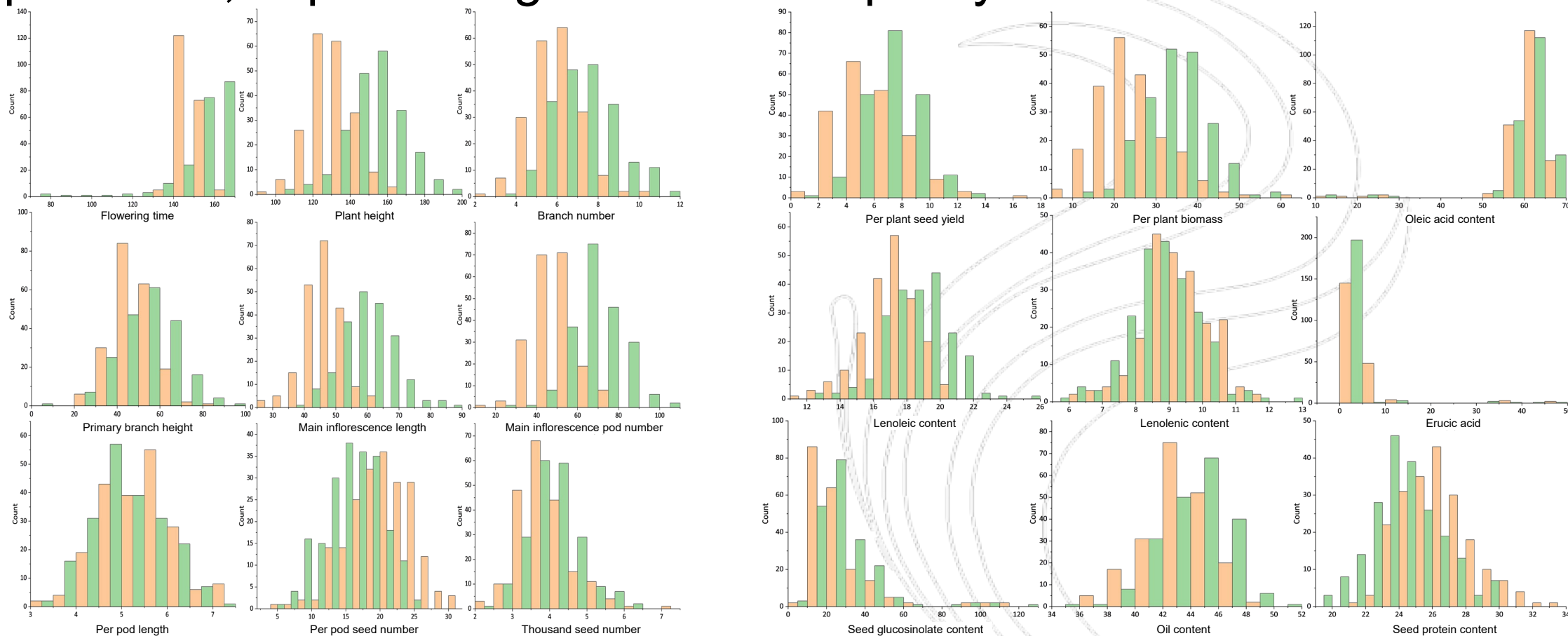
(Hu et al., unpublished data)

# Mapping more regions for SV and exotic introgressions



# Progress 2 Trait-SV/marker associations for NP

- Phenotyping on three DH populations and GWAS populations: >3 years, 3 replications, important agronomic and quality traits



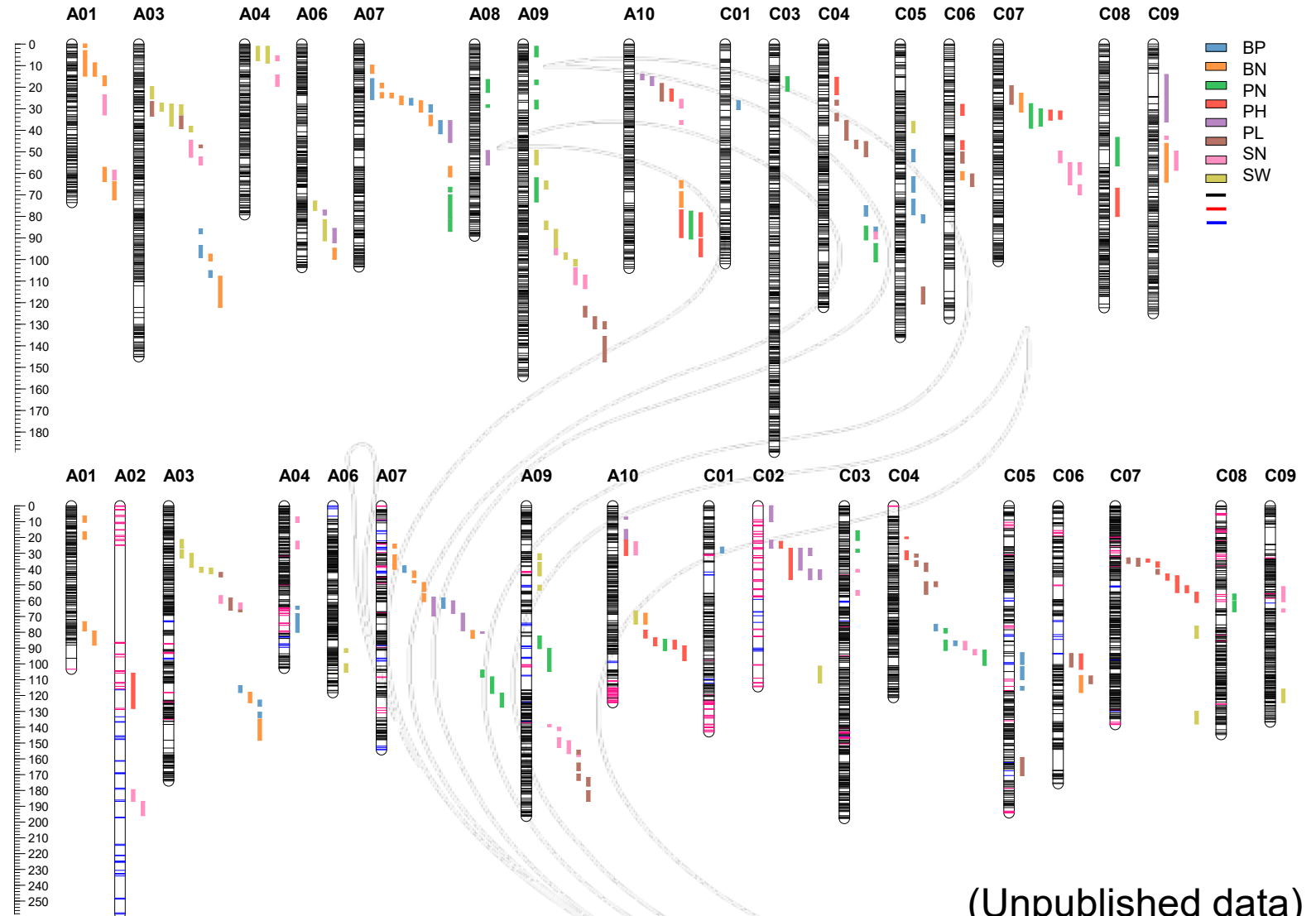
The new-type *B. napus* GWAS population has abundant phenotypic variations

(Han Qin, unpublished data)

# Hundreds of QTL mapped for important traits

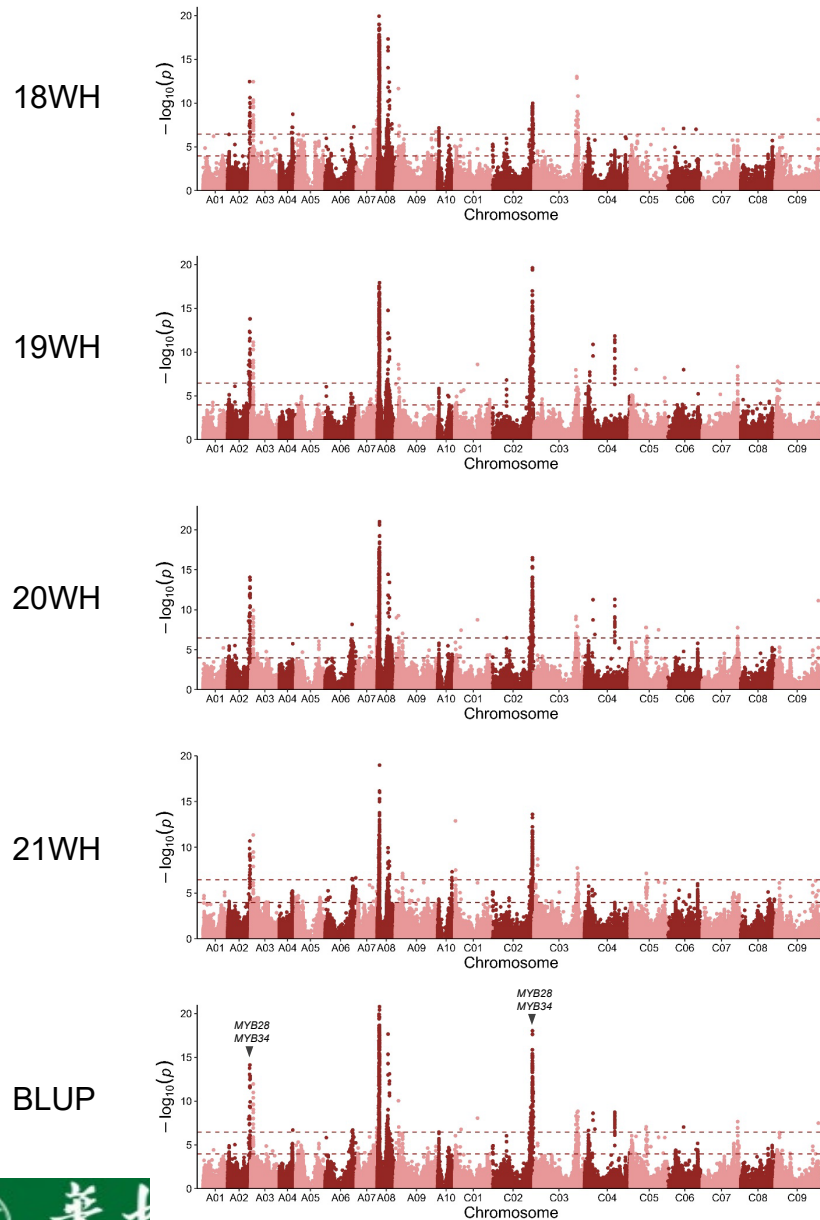
➤ ~ 500 QTL identified in the three DH mapping populations and diverse GWAS population accounting for development, yield traits, and seed quality traits

➤ ~150 possible novel QTL different with that detected in conventional *B. napus* populations



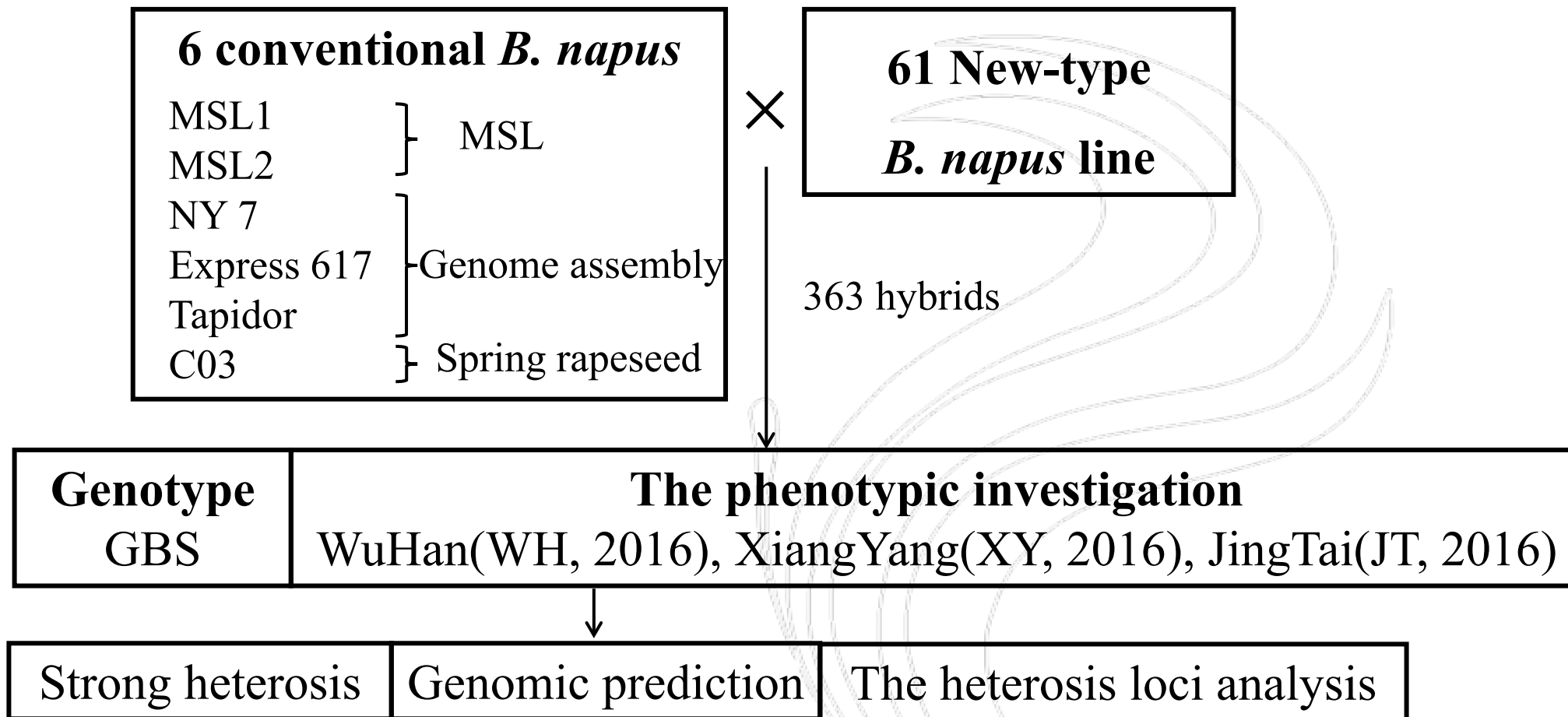
(Unpublished data)

# An example: novel QTL for seed glucosinolate content

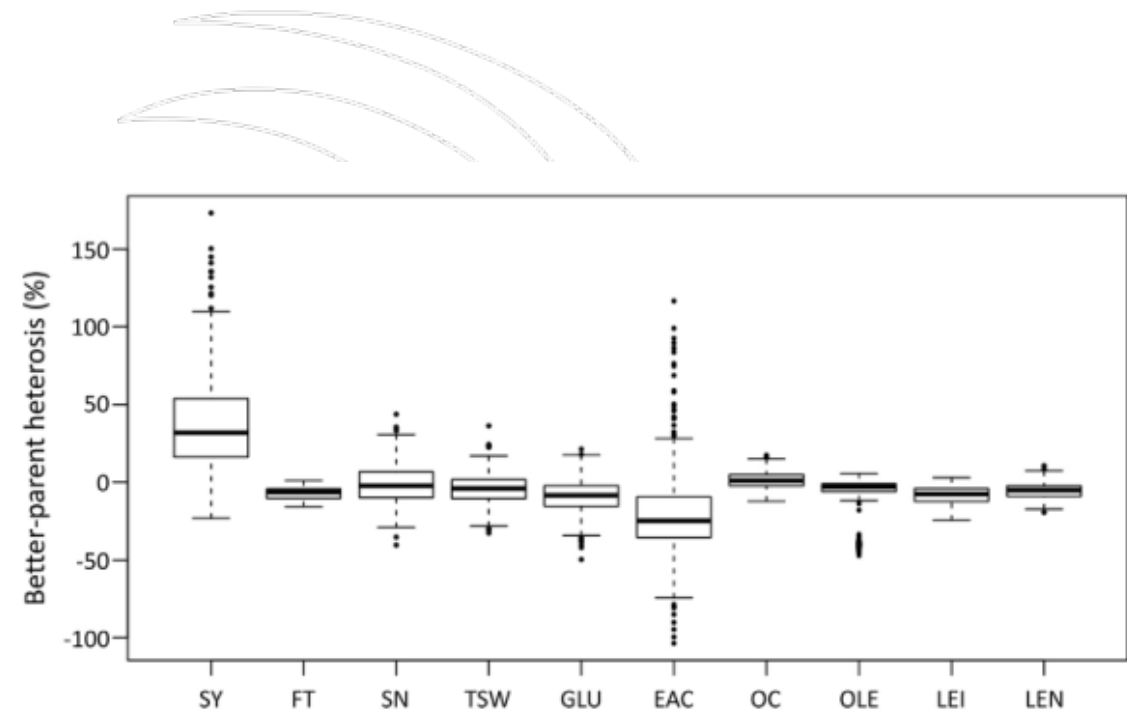
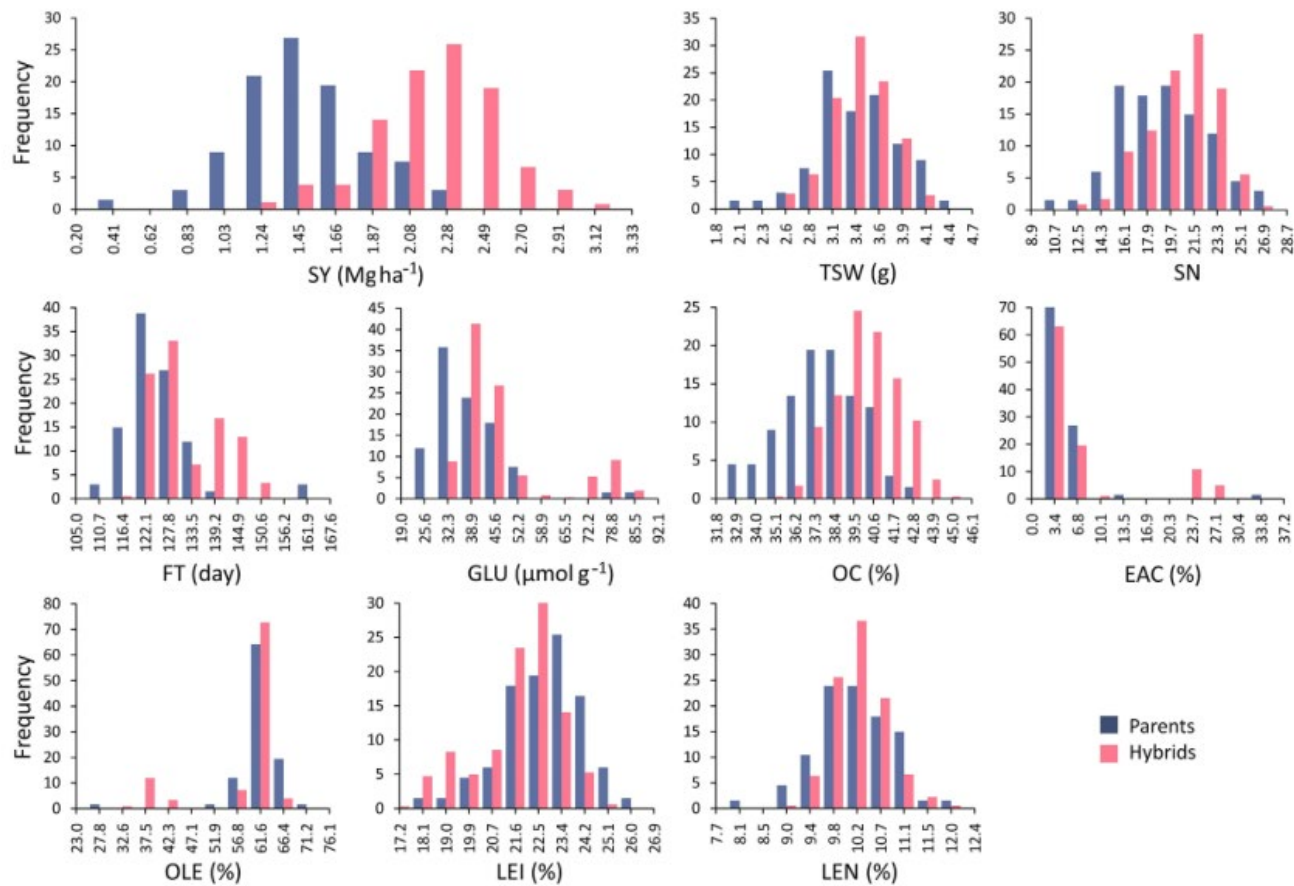


- Two major QTL were identified on chromosome A02 and C02, which is related to major genes *MYB28* and *MYB34*.
- Five possible novel QTLs were identified. Especially the QTL on A08 can explain ~60% of the phenotypic variation.

# Progress 3 Heterosis-SV/marker associations and GS



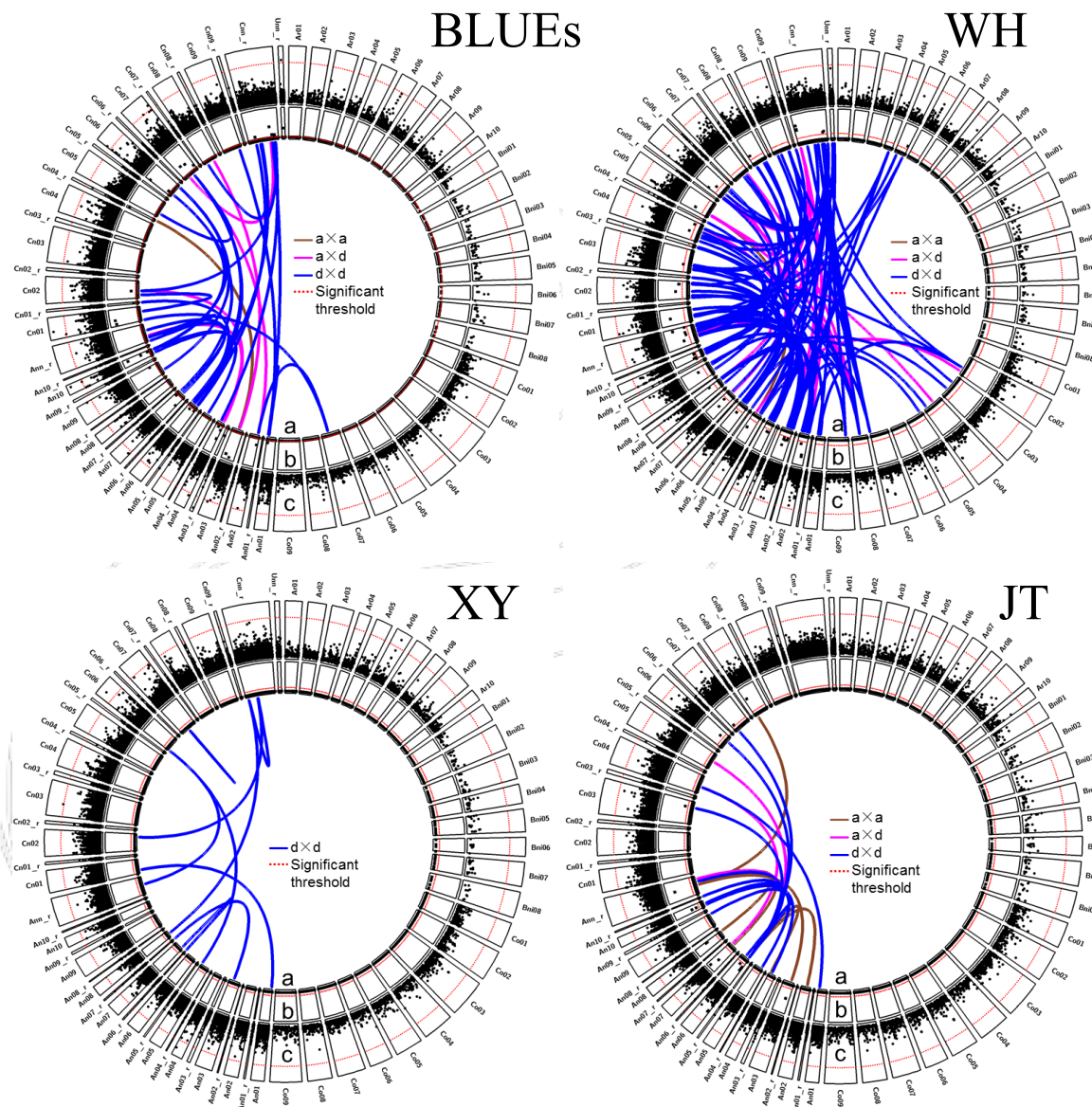
# Strong heterosis in hybrids made between new-type *B. napus* and conventional *B. napus* testers



(Hu et al., 2021, Crop Journal)

# Detection of heterosis loci for seed yield

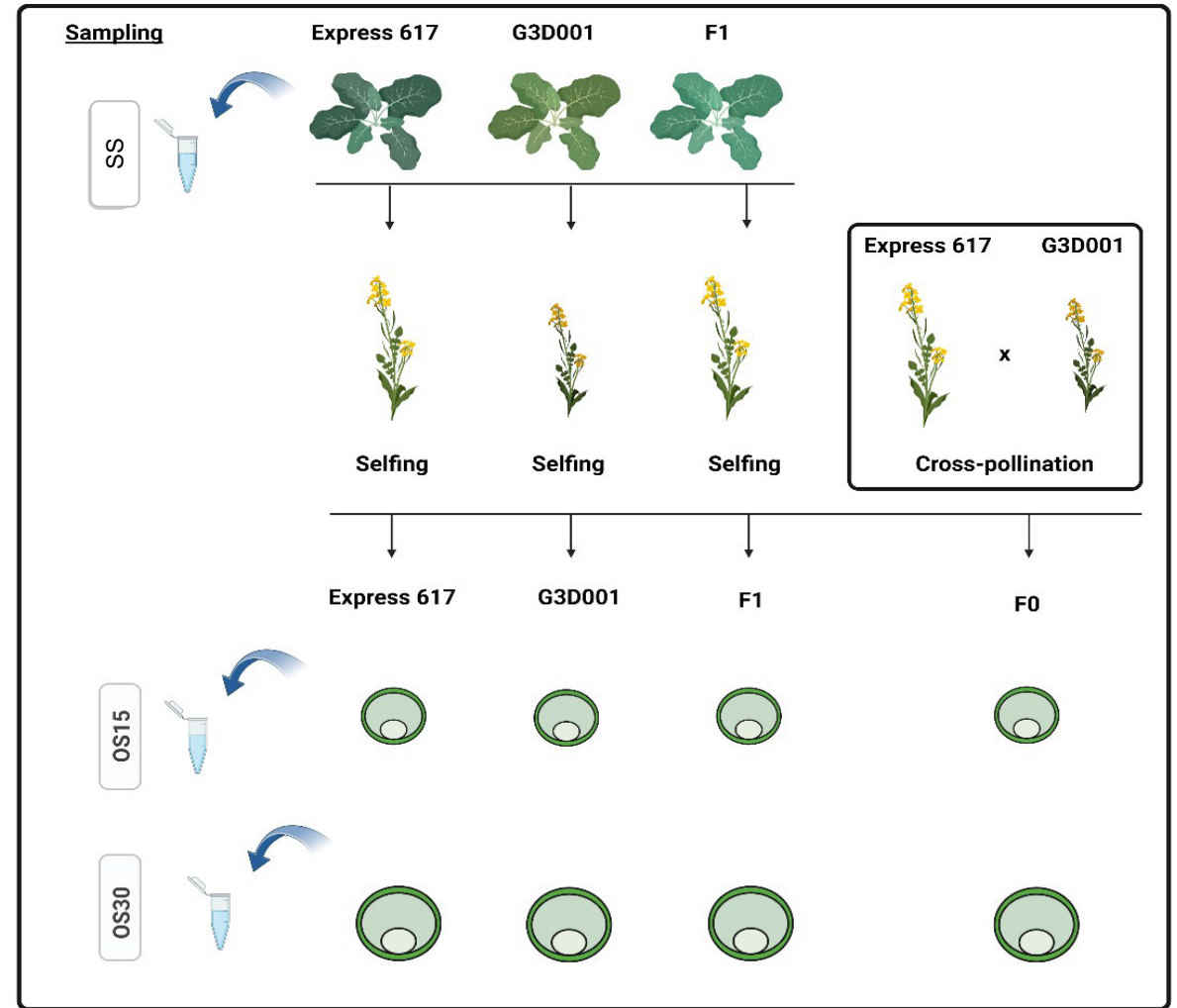
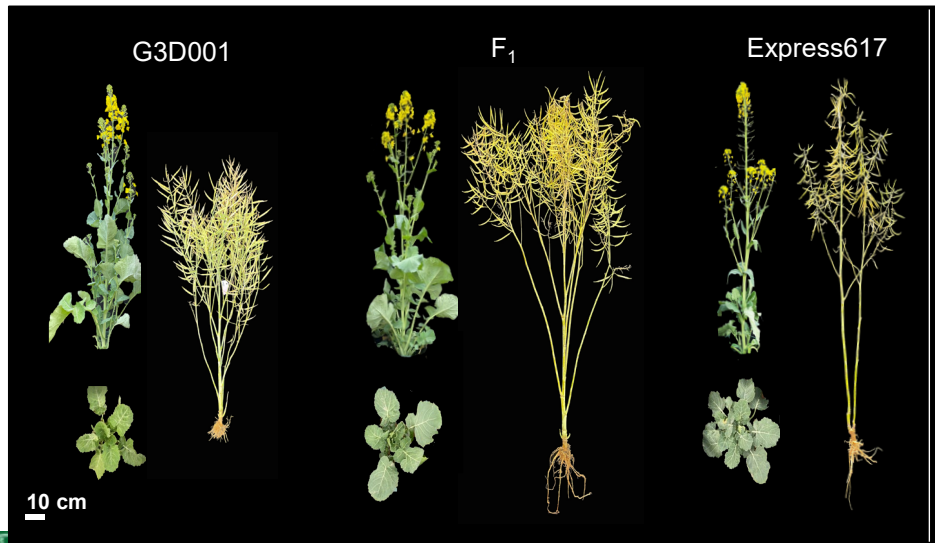
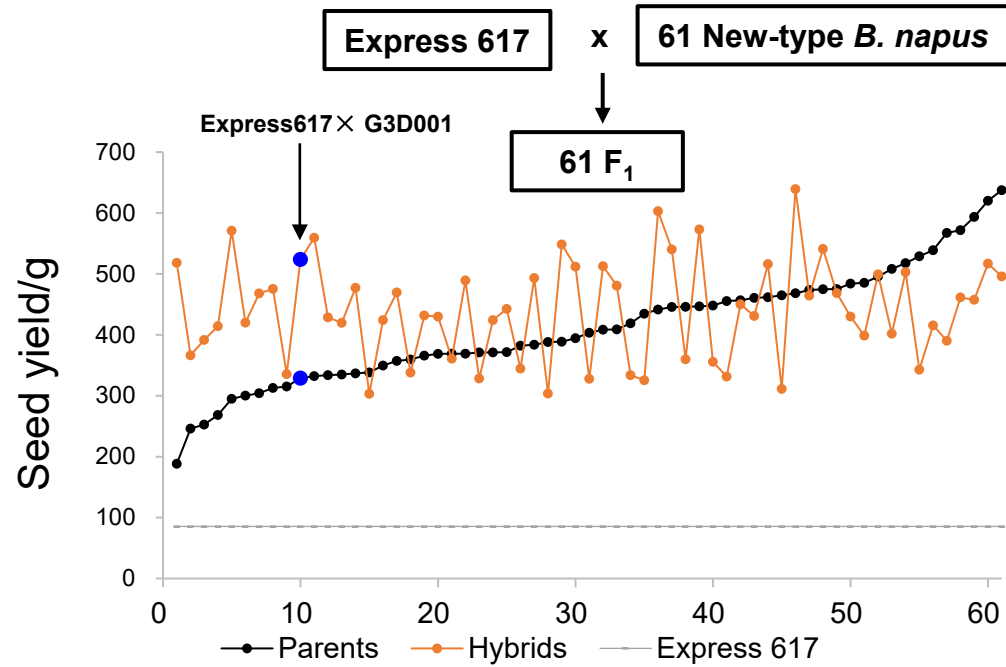
| Significant sites           | cross | WH  | XY | JT |
|-----------------------------|-------|-----|----|----|
| No. significant d effects   | 13    | 0   | 5  | 0  |
| No. significant a×a effects | 1     | 1   | 0  | 9  |
| No. significant a×d effects | 7     | 39  | 0  | 4  |
| No. significant d×d effects | 31    | 136 | 10 | 26 |
| No. heterotic effects       | 33    | 47  | 18 | 24 |



More heterosis sites and d×d and a×d effects were detected in WH environment

(Hu et al., 2021)

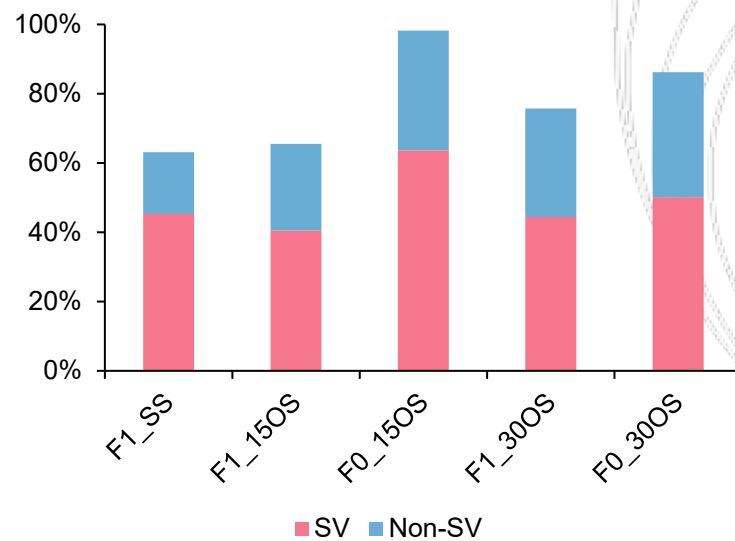
# Understanding genomic-basis of heterosis with omics-approach



Cooperation with Prof. Rod Snowdon on NSFC-DFG cooperation project (2019-2021)

# Predominant maternal dominant expression in **DEGs**

| Type | Feature     | Additivity |       | ELD-G3D001 |        | ELD-Express 617 |        | Heterotic down-regulation |       |       | Heterotic up-regulation |       |        |
|------|-------------|------------|-------|------------|--------|-----------------|--------|---------------------------|-------|-------|-------------------------|-------|--------|
|      |             | I          | XII   | II         | XI     | IV              | IX     | III                       | VII   | X     | V                       | VI    | VIII   |
|      |             | E-F-G      | E-F-G | E-F-G      | E-F-G  | E-F-G           | E-F-G  | E-F-G                     | E-F-G | E-F-G | E-F-G                   | E-F-G | E-F-G  |
|      |             |            |       |            |        |                 |        |                           |       |       |                         |       |        |
| mRNA | DEG_F1_SS   | 25.20%     | 8.68% | 9.01%      | 20.34% | 15.06%          | 18.75% | 0.08%                     | 0.77% | 0.19% | 0.13%                   | 0.18% | 1.61%  |
|      | DEG_F1_15OS | 9.79%      | 6.44% | 9.59%      | 18.16% | 13.20%          | 24.58% | 0.89%                     | 3.64% | 0.26% | 0.49%                   | 1.53% | 11.43% |
|      | DEG_F0_15OS | 0.55%      | 0.08% | 0.09%      | 1.04%  | 46.34%          | 50.78% | 0.09%                     | 0.13% | 0.00% | 0.00%                   | 0.50% | 0.40%  |
|      | DEG_F1_30OS | 8.75%      | 5.45% | 5.49%      | 11.72% | 20.69%          | 37.91% | 0.81%                     | 1.74% | 0.06% | 0.28%                   | 2.32% | 4.80%  |
|      | DEG_F0_30OS | 7.37%      | 2.18% | 1.93%      | 3.21%  | 32.68%          | 48.52% | 0.34%                     | 0.53% | 0.01% | 0.05%                   | 1.38% | 1.81%  |

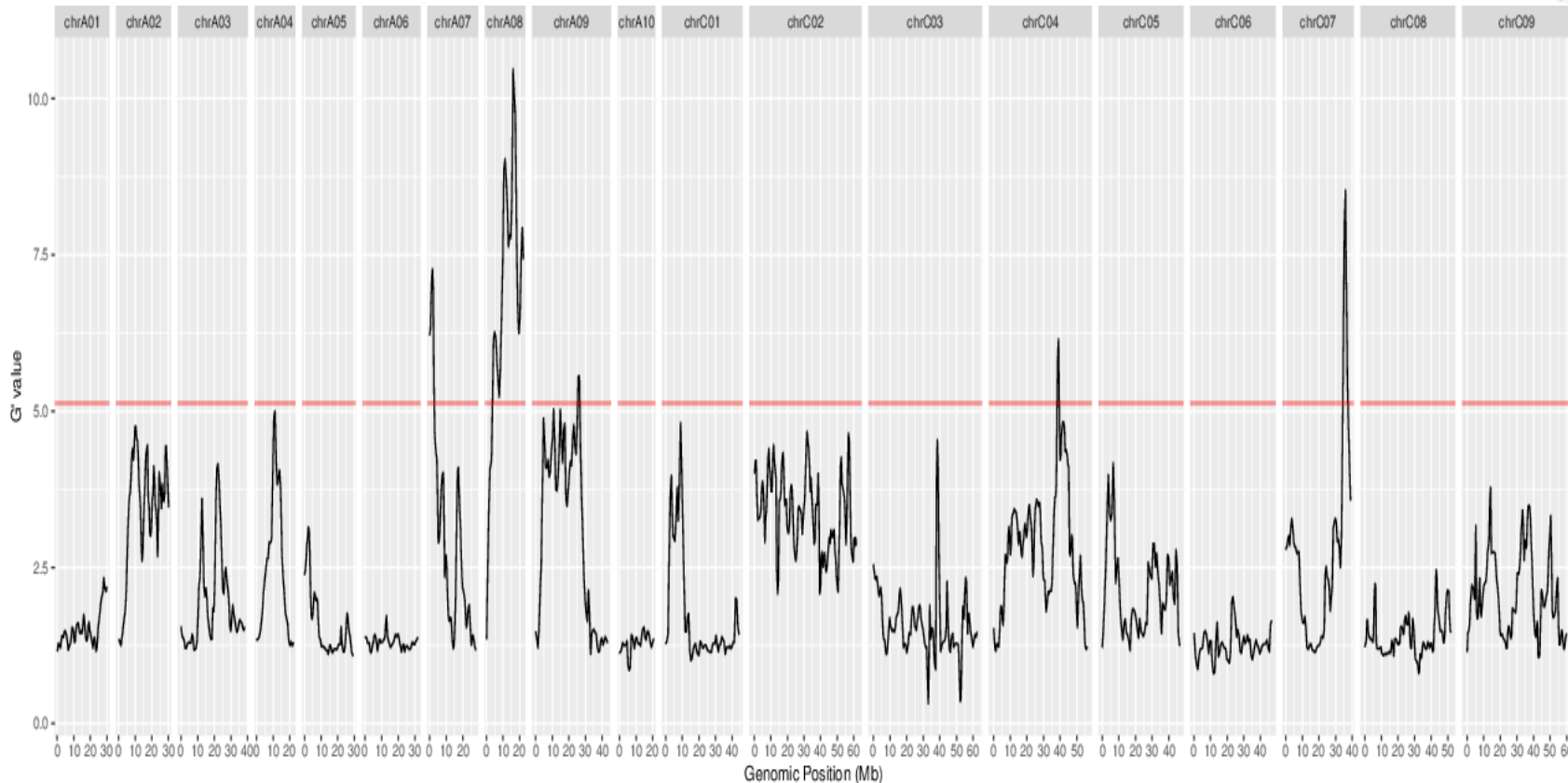


(Orantes-Bonilla et al., 2023, TAG)

The dominant expression level of SV-DEGs were higher than that of the non-SV DEGs

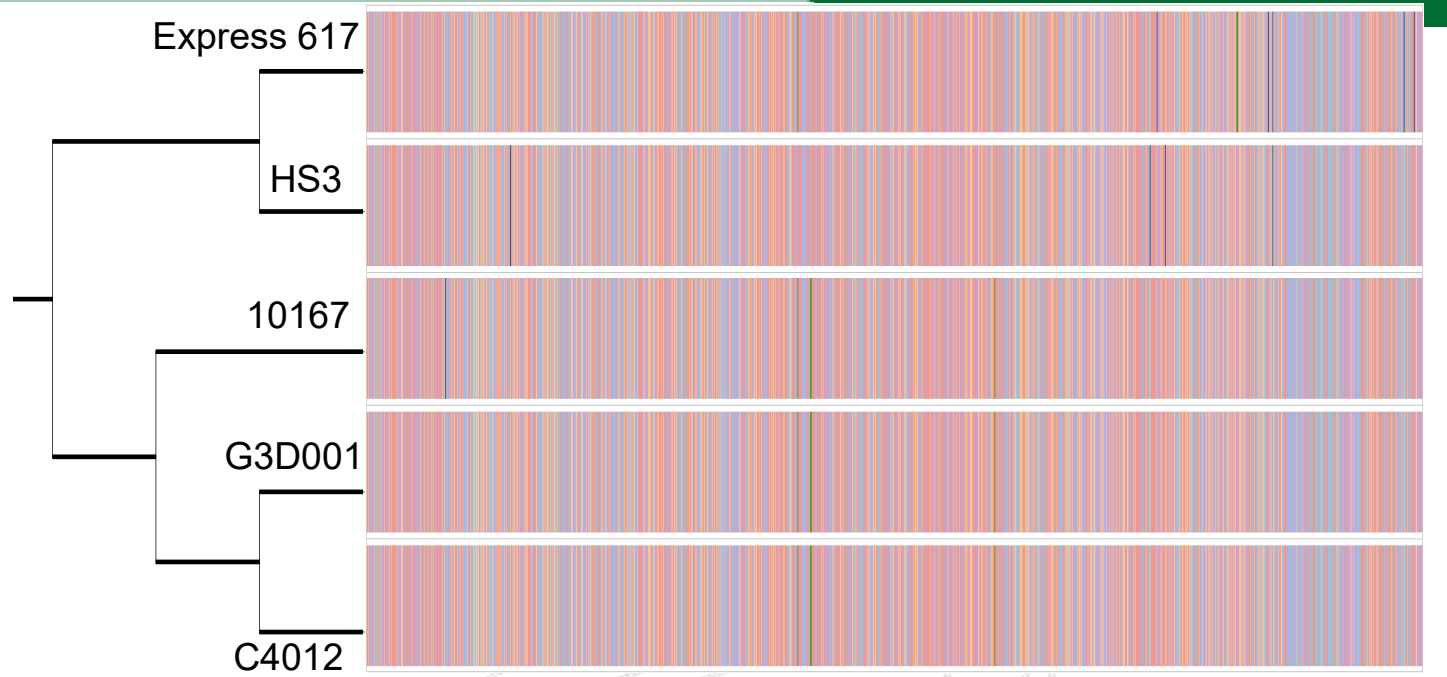
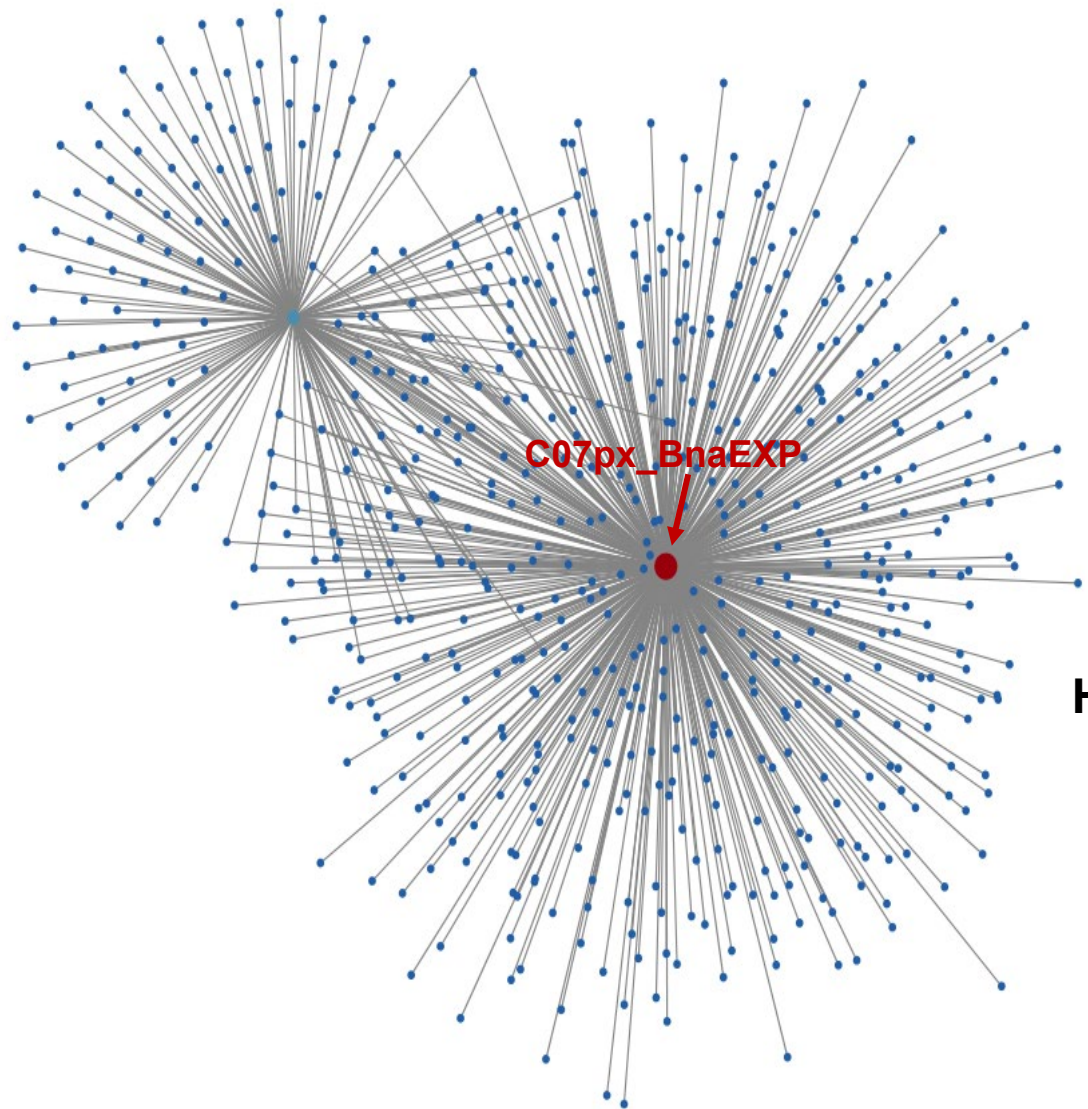
(Hao Wang, unpublished data)

# Further QTL mapping in EGF<sub>2:3</sub> population revealed an exotic introgression of *B. carinata* increasing seed weight

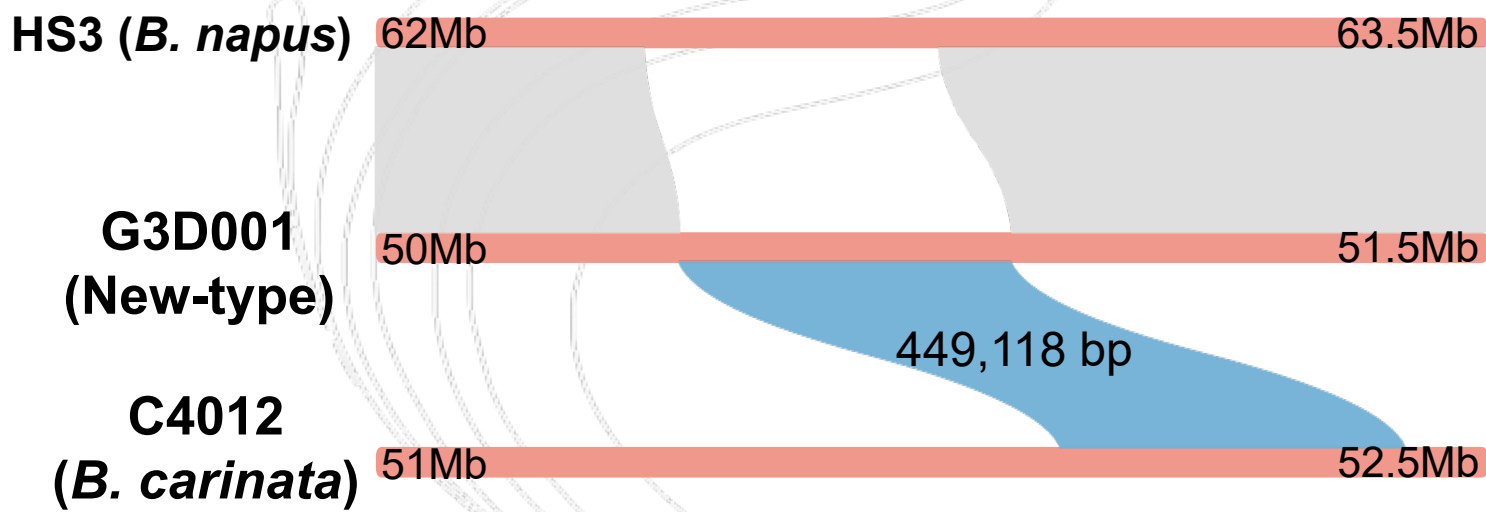


|                     |              |
|---------------------|--------------|
| <b>Chr</b>          | C07          |
| <b>Position(cM)</b> | 130          |
| <b>Effect.a</b>     | -0.0968      |
| <b>Effect.d</b>     | -0.2146      |
| <b>LOD</b>          | 5.7164       |
| <b>Left marker</b>  | 33897194     |
| <b>Right marker</b> | 35451143_pav |
| <b>R2(%)</b>        | 13.1106      |

This QTL on C07 were detected by both of the BSA and QTL mapping

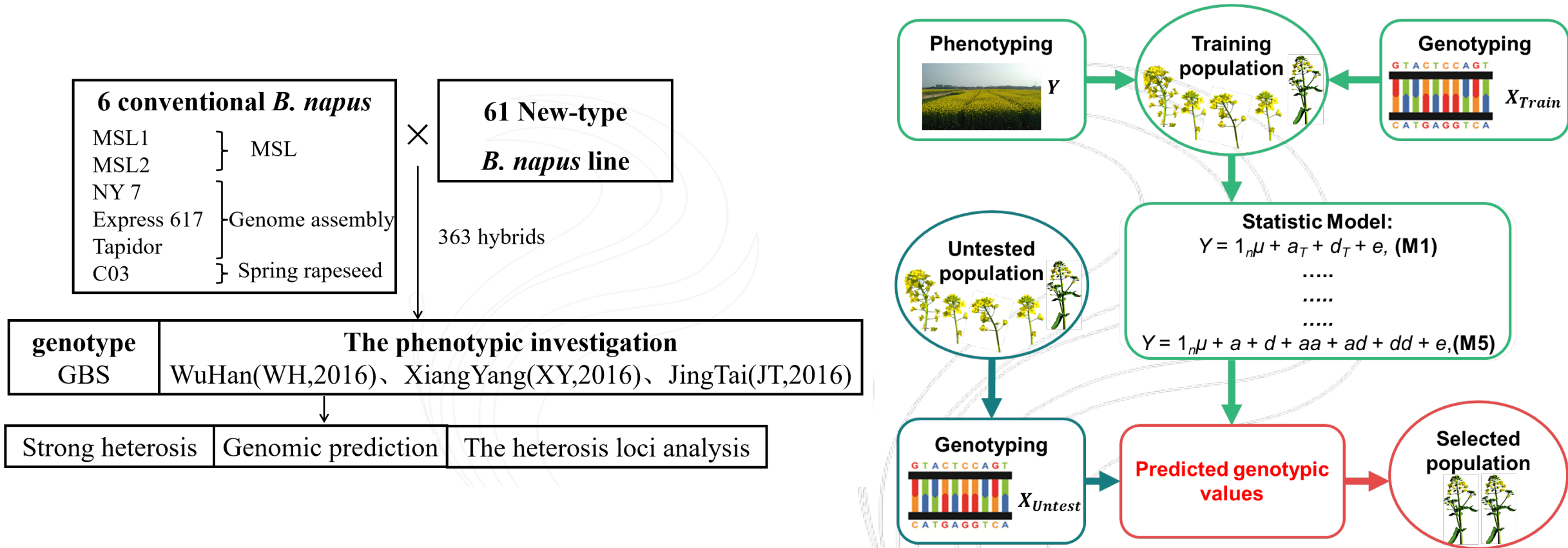


— A   
 — T   
 — C   
 — G   
 — Deletion   
 — Mismatch



(Hao Wang, unpublished data)

# GS test: using the 363 hybrids as training population



Test accuracy by constructing models incorporating different markers and genetic effects

The prediction accuracy was improved for 7 traits by adding exotic introgression-related markers

$$Y = 1_n\mu + a_T + d_T + e, \text{ (M1)}$$

$$Y = 1_n\mu + a_T + a_S + d_T + d_S + e, \text{ (M3)}$$

| Test pop. | Model | SY    | Ft    | SN    | SW    | GLSC  | EAC    | OC    | OLC    | HLC   | HLLC  |
|-----------|-------|-------|-------|-------|-------|-------|--------|-------|--------|-------|-------|
| TS2       | M1    | 0.660 | 0.758 | 0.591 | 0.589 | 0.635 | 0.691  | 0.585 | 0.626  | 0.760 | 0.669 |
|           | M2    | 0.561 | 0.742 | 0.621 | 0.585 | 0.690 | 0.711  | 0.561 | 0.677  | 0.748 | 0.693 |
|           | M3    | 0.637 | 0.759 | 0.631 | 0.604 | 0.672 | 0.718  | 0.592 | 0.631  | 0.751 | 0.682 |
|           | M4    | 0.649 | 0.768 | 0.637 | 0.609 | 0.648 | 0.717  | 0.609 | 0.656  | 0.756 | 0.692 |
| TS1       | M1    | 0.460 | 0.759 | 0.503 | 0.515 | 0.597 | 0.611  | 0.443 | 0.582  | 0.630 | 0.537 |
|           | M2    | 0.377 | 0.737 | 0.499 | 0.497 | 0.658 | 0.642  | 0.436 | 0.625  | 0.636 | 0.575 |
|           | M3    | 0.442 | 0.760 | 0.513 | 0.515 | 0.636 | 0.645  | 0.451 | 0.581  | 0.627 | 0.540 |
|           | M4    | 0.462 | 0.770 | 0.515 | 0.500 | 0.609 | 0.639  | 0.445 | 0.605  | 0.625 | 0.548 |
| TS0       | M1    | 0.173 | 0.583 | 0.179 | 0.392 | 0.076 | -0.250 | 0.106 | -0.145 | 0.142 | 0.137 |
|           | M2    | 0.006 | 0.530 | 0.129 | 0.394 | 0.223 | -0.117 | 0.069 | -0.055 | 0.212 | 0.190 |
|           | M3    | 0.127 | 0.552 | 0.176 | 0.402 | 0.104 | -0.170 | 0.107 | -0.189 | 0.184 | 0.135 |
|           | M4    | 0.129 | 0.574 | 0.221 | 0.383 | 0.025 | -0.195 | 0.095 | -0.163 | 0.150 | 0.129 |

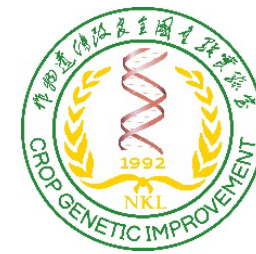
The best models for different traits varied, and the predication accuracy was the highest for multi-environment intergrated data

| Env.           | M  | SY    | FT    | SN    | SW    | GSLC  | EAC   | OC    | OLC   | HLC   | HLLC  |
|----------------|----|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| Mul-integrated | M1 | 0.728 | 0.972 | 0.751 | 0.649 | 0.971 | 0.990 | 0.690 | 0.987 | 0.910 | 0.828 |
|                | M2 | 0.685 | 0.964 | 0.769 | 0.650 | 0.964 | 0.988 | 0.693 | 0.986 | 0.906 | 0.822 |
|                | M3 | 0.727 | 0.971 | 0.765 | 0.654 | 0.970 | 0.989 | 0.700 | 0.987 | 0.911 | 0.825 |
|                | M4 | 0.734 | 0.973 | 0.754 | 0.651 | 0.971 | 0.990 | 0.701 | 0.987 | 0.913 | 0.829 |
| WH             | M1 | 0.574 | 0.959 | 0.771 | 0.569 | 0.953 | 0.980 | 0.484 | 0.975 | 0.853 | 0.746 |
|                | M2 | 0.581 | 0.952 | 0.780 | 0.591 | 0.948 | 0.978 | 0.512 | 0.973 | 0.853 | 0.722 |
|                | M3 | 0.585 | 0.959 | 0.782 | 0.586 | 0.951 | 0.979 | 0.501 | 0.974 | 0.859 | 0.738 |
|                | M4 | 0.578 | 0.961 | 0.775 | 0.569 | 0.954 | 0.980 | 0.475 | 0.975 | 0.858 | 0.745 |
| XY             | M1 | 0.347 | 0.947 | 0.692 | 0.694 | 0.980 | 0.989 | 0.426 | 0.989 | 0.912 | 0.687 |
|                | M2 | 0.245 | 0.939 | 0.702 | 0.740 | 0.978 | 0.987 | 0.424 | 0.988 | 0.913 | 0.681 |
|                | M3 | 0.318 | 0.947 | 0.705 | 0.733 | 0.980 | 0.988 | 0.441 | 0.989 | 0.915 | 0.688 |
|                | M4 | 0.337 | 0.951 | 0.699 | 0.716 | 0.979 | 0.989 | 0.420 | 0.990 | 0.912 | 0.687 |
| JT             | M1 | 0.619 | 0.849 | -     | 0.581 | 0.967 | 0.988 | 0.771 | 0.980 | 0.890 | 0.804 |
|                | M2 | 0.564 | 0.801 | -     | 0.541 | 0.963 | 0.985 | 0.735 | 0.977 | 0.881 | 0.804 |
|                | M3 | 0.614 | 0.843 | -     | 0.572 | 0.967 | 0.987 | 0.760 | 0.978 | 0.884 | 0.802 |
|                | M4 | 0.611 | 0.847 | -     | 0.584 | 0.967 | 0.987 | 0.771 | 0.980 | 0.887 | 0.803 |

# Summary

- Frequent large-scale SV drives the genome diversity and plasticity of *de novo* domesticated *B. napus*
- SVs were not randomly generated and could be fixed through influence on phenotypes during artificial selection
- Hundreds of QTL were identified in new-type *B. napus* and their derived hybrid populations
- The novel QTL identified in new-type *B. napus* population would provide valuable genetic resource for rapeseed breeding
- The GS accuracy would be further improved by adding SV-related markers
- Further genome-based approach will be performed to explore and add the breeding value of the novel gene pool

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